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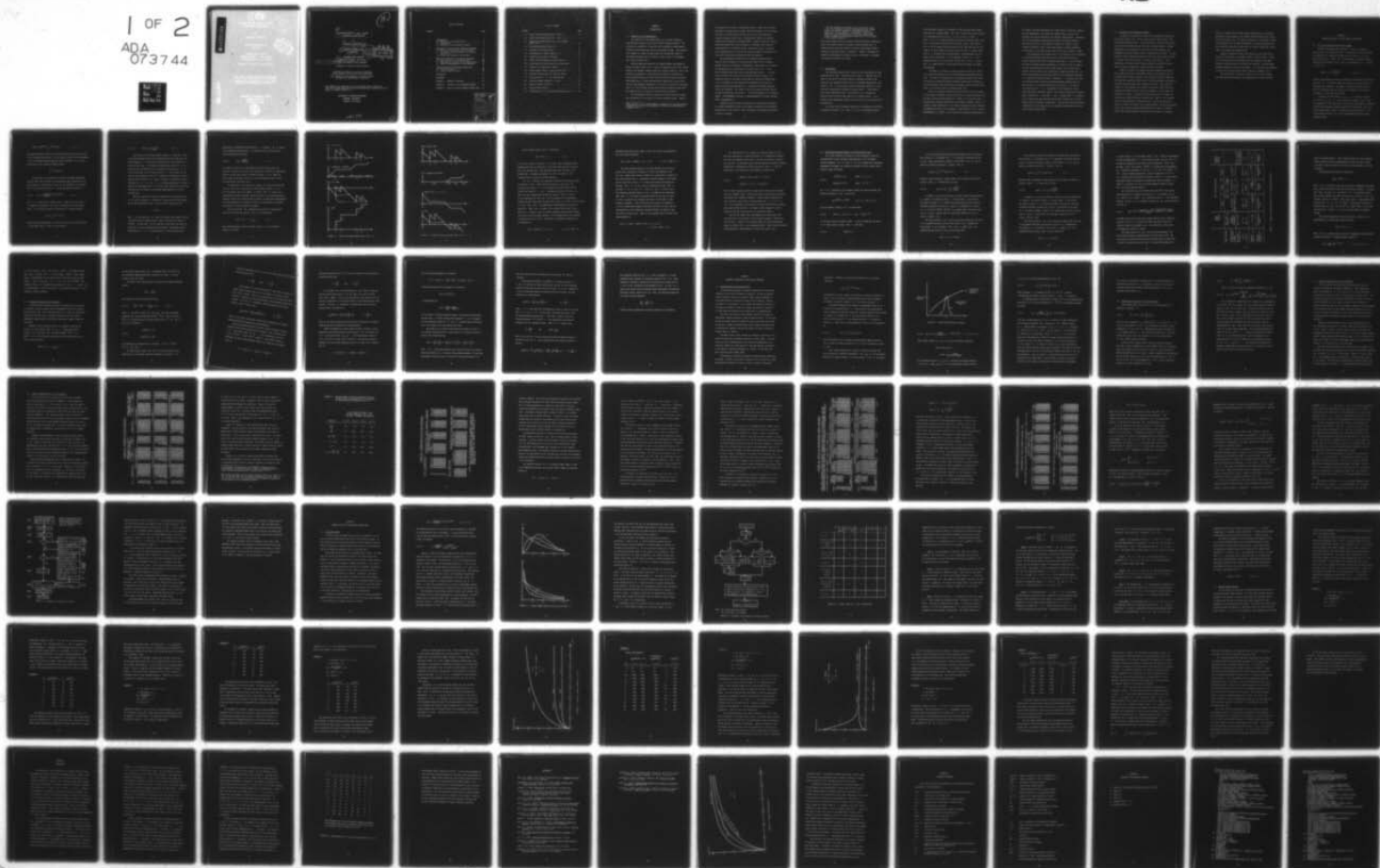
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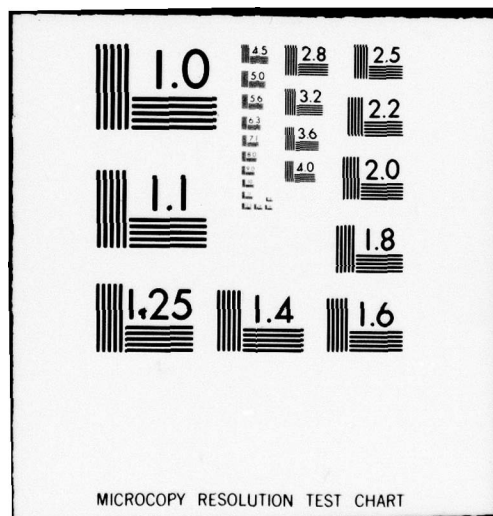
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TRANSIENT EFFECTS IN M/G/1 QUEUES:
AN EMPIRICAL INVESTIGATION,

by

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MICHAEL R. MIDDLETON

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CHAPTER 1

INTRODUCTION

1.1. Objective of the Dissertation

The objective of this dissertation is to provide tables for time-dependent expected server load in M/G/1 queueing systems.¹ The results are presented in a form that can be applied by practitioners involved in the design and control of operating systems. This presentation attempts to bridge the gap between the mathematical theory of stochastic processes and the numerical results useful to the management science practitioner.

Waiting lines and congestion are common problems encountered in almost everyone's daily life. A queue can develop in any service system whenever the immediate demand exceeds the system's capacity. One of the problems in designing or controlling such systems is achieving the proper balance between the level of service and the amount of waiting which might occur. Formal analysis may be appropriate if the service involves very expensive equipment, if the costs of waiting are extremely high, or if the decision involves many similar operating systems where the combined costs of service or waiting could be excessive.

The mathematical theory of queues can often provide some insight into the behavior of an actual or proposed operating system. Ideally,

¹This notation is the standard Kendall designation for queueing systems: Markovian (or Poisson) arrivals/General service time distribution/1 (single) server.

the designer would prefer a mathematical model or algorithm that would determine the optimal system design, given the demand or arrival pattern, the costs of various service levels, and the costs associated with customer waiting or delay times. Unfortunately, there is no general optimization technique for queueing models. However, there are numerous predictive models that allow an analyst to determine some operating characteristics of a specific system. Prospective levels of service can be analyzed one at a time and an evaluative model can determine the total cost of service and waiting for each alternative.

The mathematical formulas for such operating characteristics that are often available to the practitioner apply only to queueing systems that are in statistical equilibrium. These steady state results are appropriate for a system where the server can, on the average, serve customers faster than they arrive and where sufficient time has passed to cancel the effects of the system's initial condition. If customers arrive at the system faster than the server can handle them, or if the behavior of the system must be analyzed for the start-up period, then the time-dependent (i.e., transient) operating characteristics are important. Lee [1966, p. 26], in an opinion probably shared by many practitioners, has stated "... the first working rule of queueing theory: time-dependent solutions to queueing-models are either unobtainable or unmanageable."

Transient solutions that are available in the queueing literature are usually expressed in terms of the Laplace transform of the operating characteristic. Bhat [1969, p. B284] reiterates the problem and mentions a possible solution:

"Plainly speaking, the results, given in terms of transforms, very often with more than one argument, fail to make sense to an applied researcher. Numerical inversion of transforms ... is an answer to this problem. But at this stage, the inversion methods are either not sophisticated enough to handle the more complex situations or do not appeal to the applied researcher."

The present research uses an accurate transform inversion technique to produce numerical results for transient expected system load. A practitioner can use these tables to analyze a wide range of M/G/1 systems by referring directly to Chapter 4. Chapter 2 discusses the transform relationships from which we start, and Chapter 3 discusses the inversion technique to be used.

1.2. Methodology

The operating characteristic studied in this dissertation is time-dependent server load. The server load at epoch t , denoted $W(t)$, is equal to the sum of the service times of customers waiting in the queue plus any remaining service time of a customer being served. The quantity $W(t)$ is also called virtual waiting time, because it is the time that a hypothetical customer arriving at epoch t would have to wait before beginning service. Our objective is to tabulate the expected value of server load, $E[W(t)]$, for various epochs t and various system parameters (initial load, arrival rate, and service time distribution).

Our study of M/G/1 queueing systems can be embedded in the following general framework. Let $\{X(t), t \geq 0\}$ be a stochastic process

with stationary independent increments (a Lévy process) whose sample paths have no negative jumps. Let $W(t)$ be this same process modified by a reflecting barrier at zero. If $X(t) = S(t) - t$, where $\{S(t), t \geq 0\}$ is a compound Poisson process with positive jumps, then $W(t)$ is the server load process for an M/G/1 queue. (The jumps of $S(t)$ occur at customer arrival epochs and the jump sizes are service times.) We shall discuss two other choices for the X process which lead to W processes that provide bounds or approximations for the M/G/1 server load process. In one instance, we take $X(t)$ to be Brownian motion, and, in the other, we take $X(t) = G(t) - t$, where $G(t)$ is a gamma process (a Lévy process whose increments are gamma distributed).

In Chapter 2 we present a Laplace transform result for $E[W(t)]$ that covers both these two cases and the queueing (compound Poisson) case. In terms of its application to queueing processes, this result is valid for a system that begins operation either with or without an initial backlog of work and that has an average arrival rate less than, equal to, or greater than its average service rate.

There is no general closed-form expression for the exact inverse of the Laplace transform of mean server load, but numerical methods can be used to obtain an approximation of $E[W(t)]$ at a specific epoch t . The numerical technique employed in this research computes $E[W(t)]$ by taking a linear combination of the Laplace transform function evaluated at appropriate values of its argument. Theoretically, a more accurate approximation of $E[W(t)]$ can be obtained by using more evaluations of

the Laplace transform expression, but, when using an electronic computer for the computations, its finite word length places a limit on the accuracy that can be achieved in the $E[W(t)]$ approximation. Investigation of the technique using Laplace transforms with known inverses indicates that five or six significant digits for $E[W(t)]$ can be obtained efficiently; this is more than enough to justify confidence in the three or four digit results shown in the final $E[W(t)]$ tables.

The computational procedures used in this research could be applied to any M/G/1 queueing system whose service time distribution has a Laplace transform that can be evaluated numerically. In this research we achieve a balance between generality of results and ease of computation by concentrating on the well known class of M/G/1 queues whose service times have Erlang (or, equivalently, gamma) distributions. We denote these queueing systems $M/E_k/1$ and use the Erlang shape parameter k to describe the service times. (Parameter k is usually restricted to positive integer values when describing Erlang distributions, but here we allow k to assume any positive real value.) For example, the special case of $k = 1$ corresponds to the exponential service time distribution (an M/M/1 system). The present research also examines $M/E_k/1$ queues with k less than 1 and k greater than 1, corresponding to service time distributions with greater variance than the exponential and less variance, respectively. Most service time distributions encountered in actual practice can be adequately approximated using the very flexible Erlang family.

1.3. Overview of the Subsequent Chapters

In Chapter 2 we explain the sample path relationship between the server load process $W(t)$ and the associated net input process $X(t)$ in an M/G/1 queueing system. In this discussion, the net input process can actually be any Lévy process which has no negative jumps. Breiman [1968] gives an introduction to the theory of such processes, and Blumenthal and Gettoor [1968] provide a comprehensive treatment. We then present a result developed by Harrison [1977] for the Laplace transform of $E[W(t)]$ when the net input is a general Lévy process. This theoretically oriented chapter finally examines a scaling procedure which facilitates comparisons among the various processes.

Chapter 3 explains the Laplace inversion technique developed by Gaver [1966] which gives approximation based on the expected value of an observational density function. The accuracy of Gaver's algorithm was improved by Stehfest [1970], and we test the technique using Laplace transform functions whose exact inverses are known. These numerical results are also compared with results from Veillon [1974] which were obtained using other Laplace inversion techniques. We then apply the technique to $E[W(t)]$ in $M/E_k/1$ queueing systems, and we compare our results with Coleman [1975] who obtained exact $E[W(t)]$ for the M/M/1 case by evaluating sums of Bessel functions. After some additional checks to ensure the accuracy of our approximations, we provide documentation of the computer programs which generate the tables of $E[W(t)]$. Gaver [1966, 1968] presented limited numerical results for transient

$E[W(t)]$ in several M/G/1 queues, thereby demonstrating the potential usefulness of his technique, and Coleman [1975] presented exact results only for the M/M/1 case. The major contribution of the present research is the extensive set of tables in Appendix C, providing transient mean server load in queues with a wide range of traffic intensities, initial loads, and service time distributions.

Chapter 4 explains how the scaled results in the tables can be used by a practitioner to determine $E[W(t)]$ in M/G/1 queues. Charts of the Erlang service time distributions are presented, and simple methods of interpolation in the tables are explained. Several sample problems demonstrate the entire procedure. It is intended that Chapter 4 can be used by a practitioner without recourse to Chapters 2 or 3.

The dissertation concludes in Chapter 5 with a brief summary, some qualitative observations, and suggestions for further research.

CHAPTER 2

THEORETICAL RESULTS FOR THE SERVER LOAD PROCESS

2.1. The Server Load Process in M/G/1 Queues

The M/G/1 queueing system consists of a group of customers, a waiting room, and a service facility. We assume that customer arrivals are described by a stationary Poisson process $\{A(t); t \geq 0\}$ with mean arrival rate λ , where $A(t)$ is the number of customers arriving during the time interval $[0, t]$. Thus, the probability of n arrivals in $[0, t]$ is

$$P\{A(t) = n\} = \frac{e^{-\lambda t} (\lambda t)^n}{n!}, \quad n = 0, 1, \dots,$$

and the times between consecutive arrivals are exponentially distributed with mean $1/\lambda$. We further assume that there are no bulk arrivals, no reneging, and no balking. The number of potential customers and the size of the waiting room are assumed to be unlimited, and the queue discipline is first-come-first-served.

Customers arrive at epochs $\{t_1, t_2, \dots\}$, each with a demand for service $\{S_1, S_2, \dots\}$. These individual customer service times are independent of the interarrival times and are independent, identically distributed non-negative random variables with finite mean $E(S)$ and finite second moment $E(S^2)$. We assume that random variable S has a distribution function $F(\cdot)$ with corresponding Laplace-Stieltjes Transform (LST):

$$F^*(s) = E(e^{-sS}) = \int_0^{\infty} e^{-sx} dF(x) , \quad 0 < s < \infty .$$

The general approach used in this research requires only that this LST can be evaluated numerically. For the specific cases to be investigated, S has a continuous distribution with density function $f(\cdot)$, so the LST reduces to the ordinary Riemann integral

$$\int_0^{\infty} e^{-sx} f(x) dx .$$

In particular, we examine M/G/1 systems with gamma distributed service times, but we employ the terminology usually reserved for the Erlang family of distributions, where the two parameters are the mean $E(S)$ and the shape parameter k . The Erlang density function is

$$(2.1.1) \quad f(x) = \frac{[k/E(S)]^k}{(k-1)!} x^{k-1} e^{-kx/E(S)} , \quad x \geq 0 ,$$

with $k \geq 1$ restricted to integer values. However, here we allow k to assume any non-negative real value, requiring that the factorial $(k-1)!$ in the density function be replaced by the gamma function,

$$\Gamma(k) = \int_0^{\infty} x^{k-1} e^{-x} dx .$$

Using the Erlang terminology, the LST of our service time distribution is (see Drake [1967, p. 138] for a derivation)

$$(2.1.2) \quad F^*(s) = \left[\frac{k}{k + sE(S)} \right]^k, \quad k \geq 0,$$

The variance of an Erlang random variable is $[E(S)]^2/k$. Thus, for various Erlang service time distributions with the same mean, the shape parameter k is inversely proportional to the variability of those service times. For example, Erlang service times with $0 < k < 1$ have even more variability than an exponential distribution ($k = 1$). On the other hand, for very large values of k the variance is very small, and we approach the case of a constant or deterministic service time (an M/D/1 system) as k tends to infinity. Probability density functions for $0 < k \leq 1$ and $k \geq 1$ are shown graphically in Figure 4.1. Hillier and Lieberman [1974, p. 417] state that "empirical service-time distributions can usually be reasonably approximated by an Erlang distribution."

The most important descriptive parameter for a queueing system is its traffic intensity ρ , defined as the mean service time divided by the mean interarrival time. Thus, for M/G/1 systems, we have

$$\rho = \lambda E(S) .$$

When ρ is less than one, i.e., when the average time between arrivals is greater than the average service time, we say that the system is "stable". In this case, ρ is the fraction of time that the server is busy, and it can be interpreted as the system's "utilization factor". Furthermore, for $\rho < 1$ the distribution of virtual waiting time

approaches an equilibrium distribution as t increases. Let W denote this steady-state waiting time. Its expected value is given by the Pollaczek-Khintchine formula,

$$(2.1.3) \quad E(W) = \frac{\lambda E(S^2)}{2(1-\rho)}, \quad \rho < 1.$$

Our numerical results for time-dependent virtual waiting time will allow us to observe how quickly this steady-state condition is approached. However, we will also examine "unstable" systems ($\rho \geq 1$), where the queue length and waiting time tend to increase without bound, so that no steady-state conditions exist.

As mentioned in the introductory chapter, the virtual waiting time or server load, $W(t)$, represents the work backlog at epoch t , i.e., the accumulated, unserved demand. We define $W(t)$ in terms of the underlying work input processes, which allows us to use existing Laplace transform results for reflected Lévy processes. Sample path relationships are shown graphically in Figures 2.1 and 2.2, illustrating one possible realization for these stochastic processes.

The input process $S(t)$ represents the amount of customer work that arrives during the interval $[0, t]$ and is defined by

$$S(t) = S_1 + \cdots + S_{A(t)}, \quad t \geq 0.$$

This compound Poisson process has mean $E[S(t)] = \rho t$ and variance $\text{Var}[S(t)] = \lambda E(S^2)t$.

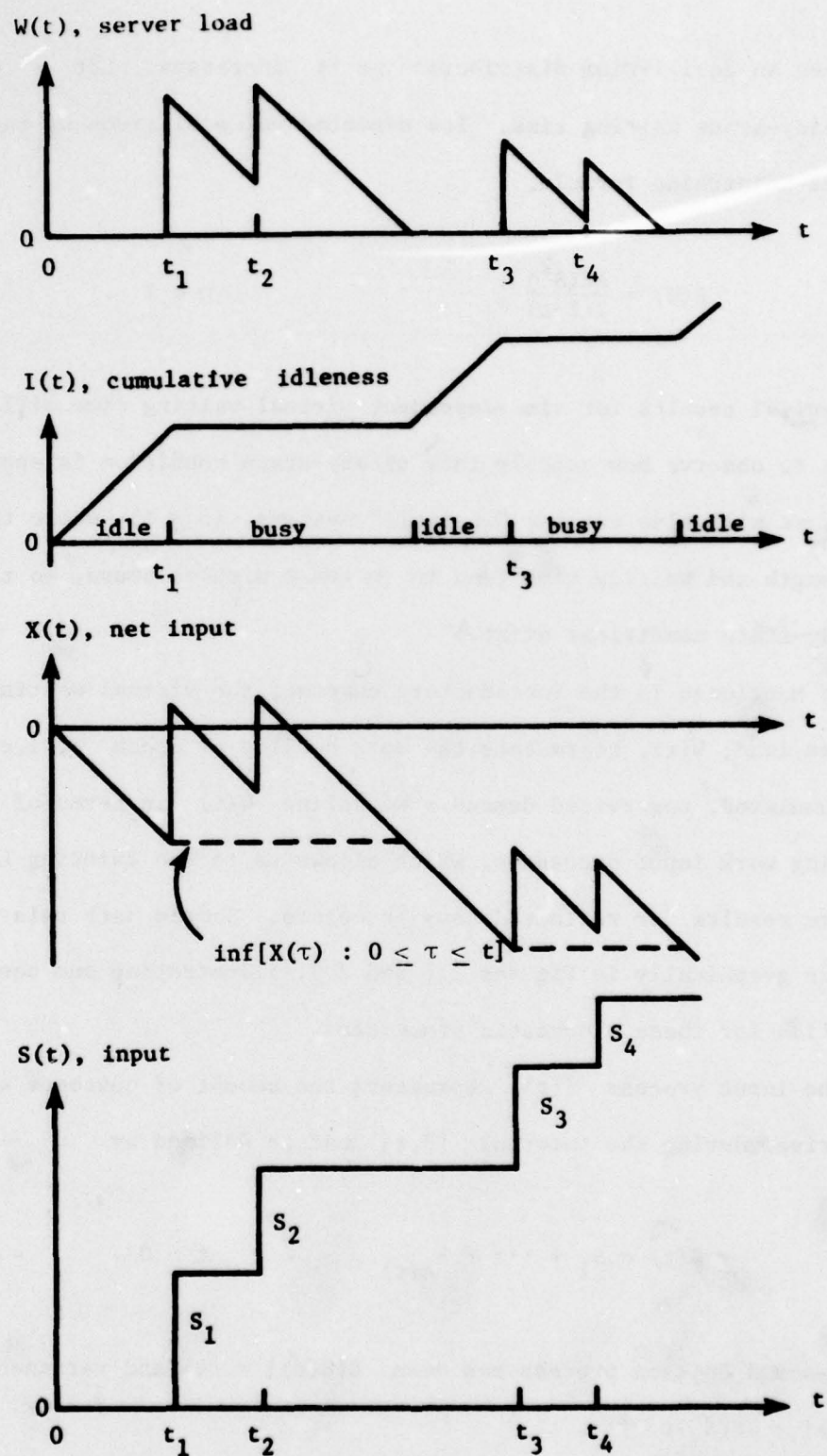


FIGURE 2.1. Sample Path Relationships when $W(0) = 0$.

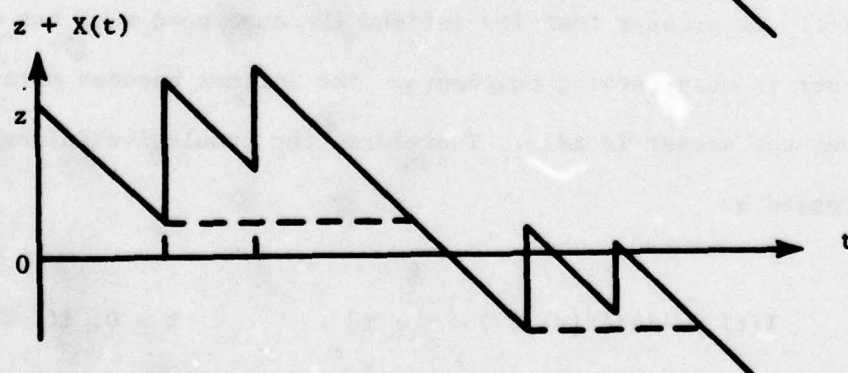
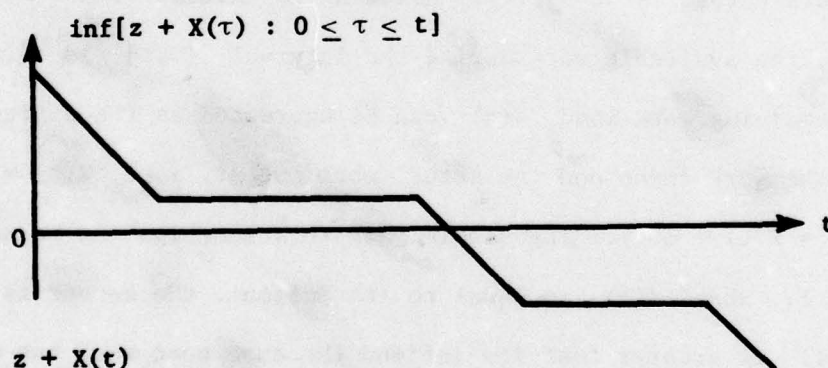
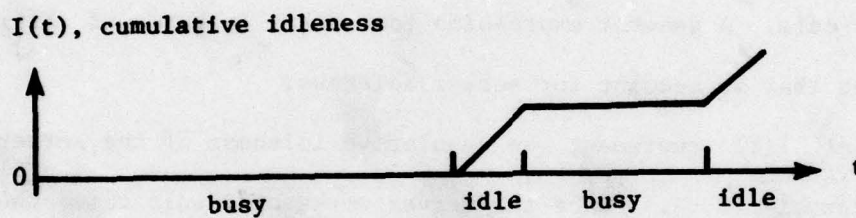
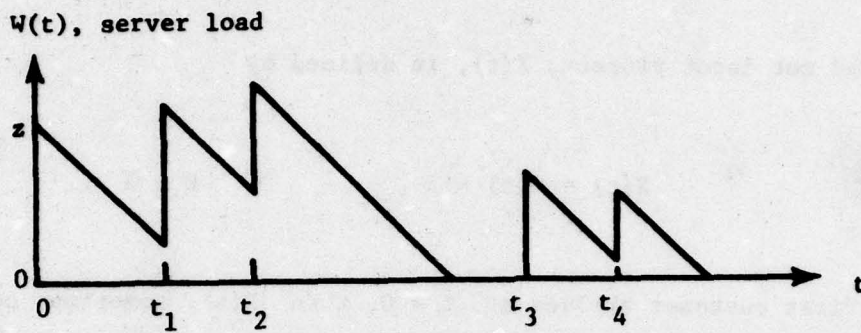


FIGURE 2.2. Sample Path Relationships when $W(0) = z = 0$.

The net input process, $X(t)$, is defined by

$$X(t) = S(t) - t, \quad t \geq 0.$$

If the first customer arrives at $t = 0$, then $X(t)$, sometimes called pseudo-server load, is identical to the server load process as long as the server remains busy. The similarity ends when the server first becomes idle. A general expression for $W(t)$ in terms of $X(t)$ requires that we account for server idleness.

Let $I(t)$ represent the cumulative idleness of the server during the interval $[0, t]$. Since the server works at a unit rate, the potential work output is t units during the same interval, and the actual work output is $t - I(t)$. Assuming no initial work load, i.e., $W(0) = 0$, the available work during the interval $[0, t]$ is $S(t)$, so the remaining work load $W(t)$ can be expressed as the difference between the work input and the actual work output, i.e., $W(t) = S(t) - [t - I(t)]$, or $W(t) = X(t) + I(t)$. The sample path relationships can be seen in Figure 2.1. When $X(t)$ is equal to its infimum, the server is idle. When $X(t)$ is greater than its infimum (because some work has arrived), the server is busy serving customers. The infimum becomes more negative only when the server is idle. Therefore, the cumulative idleness can be expressed as

$$I(t) = -\inf[X(\tau) : 0 \leq \tau \leq t], \quad t \geq 0, \text{ if } W(0) = 0.$$

Combining this with the net input, we have the server load representation (no initial workload),

$$W(t) = X(t) - \inf[X(\tau) : 0 \leq \tau \leq t] , \quad t \geq 0, \text{ if } W(0) = 0 .$$

The same reasoning applies to the more general case of initial server load, illustrated in Figure 2.2, where the underlying $S(t)$ and $X(t)$ sample paths would be identical to those shown in Figure 2.1. The initial server load z represents a specific amount of work available for service at epoch $t = 0$. Then the available work during an interval $[0, t]$ is $z + S(t)$, and the remaining work load $W(t)$ at any epoch t is $z + S(t) - [t - I(t)]$, or $W(t) = z + X(t) + I(t)$. The cumulative idleness function $I(t)$ is slightly more complicated when $W(0) > 0$. The server is initially busy in this case, so when $z + X(t)$ is equal to its infimum, the server is idle only if that infimum is negative. After the initial busy period, the infimum of $z + X(t)$ becomes more negative only when the server is idle, and the representation of the cumulative idleness function is similar to the case with no initial load. Thus, for this general case the server load representation is

$$W(t) = z + X(t) - \{\inf[z + X(\tau) : 0 \leq \tau \leq t]\}^- ,$$

$$t \geq 0, \text{ if } W(0) = z \geq 0 .$$

Our main objective is to compute the expected value of $W(t)$, where the expectation is taken with respect to a probability distribution over all possible sample paths. We adopt the notation $E_z[W(t)]$ and $E_z[I(t)]$ for expected server load and expected cumulative idleness, respectively, conditional on initial work load $W(0) = z$. From our discussion of the sample path relationships it follows that

$$E_z[W(t)] = E_z[z + S(t) - t + I(t)] ,$$

(2.1.3)

$$E_z[W(t)] = z + \rho t - t + E_z[I(t)] .$$

Thus, in M/G/1 queueing systems the mean server load can be separated into four additive terms: initial work load, new work input, potential work output, and cumulative idleness. Another useful observation is that $E_z[I(t)]$ must be zero in the interval from $t = 0$ to $t = z$, i.e., it is impossible for the server to become idle before the initial work load has been serviced. In Chapter 3 this property is used to provide a check on the accuracy of our numerical results.

Thus far the sample path relationships and expectations have been discussed in the context of M/G/1 queueing systems. However, we also calculate $E_z[W(t)]$ for processes where the same relationships apply, but where $S(t)$ is not compound Poisson. These related processes provide bounds or approximations for M/G/1 mean server load.

2.2. The Laplace Transform Result for Expected Server Load

All of the stochastic processes for which numerical values are calculated here can be discussed simultaneously in the following unified framework. Let $X = \{X(t), t \geq 0\}$ be a process with stationary, independent increments (an infinitely divisible or Lévy process) and no negative jumps, and define

$$\begin{aligned} -E[X(t)] &= \mu t, & \text{where } -\infty < \mu < \infty, \\ (2.2.1) \quad \text{Var}[X(t)] &= \sigma^2 t, & \text{where } 0 < \sigma^2 < \infty, \end{aligned}$$

for $t \geq 0$. According to the standard results for Lévy processes, the Laplace transform of $X(t)$ has the form

$$E[e^{-sX(t)}] = e^{-\Phi(s)t} \quad \text{for } s > 0 \text{ and } t \geq 0,$$

and the exponent function $\Phi(\cdot)$ is convex with

$$(2.2.2) \quad \Phi(0) = 0, \quad \Phi'(0) = \mu, \quad \text{and} \quad \Phi''(0) = \sigma^2,$$

cf. Harrison [1977] and Takacs [1967]. It can be shown that for each $s > 0$ there exists a unique $\omega(s) > 0$ such that

$$(2.2.3) \quad \Phi[\omega(s)] = s.$$

In the previous section we discussed the sample path relationships where process W is obtained from X by imposing a reflecting barrier at zero. Using the notation $E_z[W(t)] = E[W(t) | W(0) = z]$, let $P_W(z, s)$ denote the Laplace transform of $E_z[W(t)]$, that is,

$$P_W(z, s) = \int_0^{\infty} e^{-st} E_z[W(t)] dt, \quad s > 0.$$

Harrison [1977] developed a simple formula for the Laplace transform of $E_z[W(t)]$, where μ is unrestricted in sign:

$$(2.2.4) \quad P_W(z, s) = \frac{z}{s} - \frac{\mu}{s^2} + \frac{e^{-\omega(s)z}}{s\omega(s)}.$$

In Chapter 3 we use this formula to obtain accurate approximations of $E_z[W(t)]$ for specific epochs t , initial loads z , and various net input processes X . To achieve the desired accuracy the numerical inversion technique requires that $P_W(z, s)$ be computed for 34 values of s in order to obtain $E_z[W(t)]$ at a single epoch, and each evaluation of $P_W(z, s)$ requires that the functional equation (2.2.3) be solved to obtain $\omega(s)$.

The ease with which $\omega(s)$ can be calculated depends upon the form of the exponent function, and the exact form of $\phi(\cdot)$ depends upon the process X . If we specify $X(t) = S(t) - t$, where $S(t)$ is a compound Poisson process, then it can be shown that

$$\phi(s) = s - \lambda[1 - F^*(s)],$$

In the previous section we discussed the sample path relationships where process W is obtained from X by imposing a reflecting barrier at zero. Using the notation $E_z[W(t)] = E[W(t) | W(0) = z]$, let $P_W(z, s)$ denote the Laplace transform of $E_z[W(t)]$, that is,

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$$\phi(s) = s - \lambda[1 - F^*(s)],$$

cf. Prabhu [1965, p. 70] and Takacs [1967, p. 59]. Using the terminology of M/G/1 queueing theory λ is the average arrival rate and $F^*(\cdot)$ is the Laplace transform of the service time distribution. Since $S(t)$ has mean $\lambda E(S)t$ and variance $\lambda E(S^2)t$, it follows that the parameters of X defined by equations (2.2.1) are $\mu = 1-\rho$ and $\sigma^2 = \lambda E(S^2)$. If $F^*(\cdot)$ can be evaluated numerically, then the properties (2.2.2) of $\Phi(\cdot)$ indicate that the functional equation $\Phi[\omega(s)] = s$ can be solved efficiently using an elementary one-dimensional search technique. In Chapter 3 we discuss our use of the Newton-Raphson method to obtain $\omega(s)$ for $M/E_k/1$ queueing systems.

For the special case of an M/M/1 system the LST of the service time distribution is obtained by setting $k = 1$ in equation (2.1.2), i.e., $F^*(s) = 1/[1 + sE(S)]$. For a specified value of s the functional equation $\Phi[\omega(s)] = s$ is a quadratic function of $\omega(s)$, and the positive solution is

$$(2.2.5) \quad \omega(s) = \frac{-\{1 - \lambda[s+E(S)]\} + \sqrt{\{1 - \lambda[s+E(S)]\}^2 + 4sE(S)}}{2E(S)} .$$

The M/M/1 system is the only queue studied here that has an analytic solution to (2.2.3). In general the M/D/1 and $M/E_k/1$ cases will require a search to determine $\omega(s)$. The formulas for these cases are summarized in Table 2.3 below.

Two other choices for $X(t)$ yield reflected processes $W(t)$ which provide bounds or approximations for M/G/1 server load. The first case is the Wiener process, which has been used as a model for a variety of physical processes since its original development as a

	Gamma Input Process	M/G/1 Queueing Systems			Brownian Motion Process
		M/E _k /1	M/M/1	M/D/1	
Input Process	G(t) with parameters α, β	$S(t) = S_1 + \dots + S_A(t)$ where A(t) is Poisson			-
Characteristics of the Input Process	gamma distributed increments	S_i Erlang with shape parameter k	S_i exponential (k = 1)	S_i constant (k $\rightarrow \infty$)	-
Net Input Process X(t)	$X_G(t) = G(t) - t$	$X(t) = S(t) - t$			$X_W(t) = \sigma \xi(t) - \mu t$ (ξ is standard Wiener process)
$E[X(t)]$	$(\frac{\alpha}{\beta} - 1)t$	$(\lambda E(S) - 1)t$	$(\lambda E(S) - 1)t$	$(\lambda E(S) - 1)t$	$-\mu t$
$Var[X(t)]$	$(\frac{\alpha}{\beta^2} - 1)t$	$\lambda \frac{k+1}{k} E(S)^2 t$	$\lambda^2 E(S)^2 t$	$\lambda E(S)^2 t$	$\sigma^2 t$
Exponent Function $\hat{\phi}(s)$	$s - \alpha \ln(1 + \frac{s}{\beta})$	$s - \lambda \{1 - [\frac{k}{k+sE(S)}]^k\}$	$s - \frac{sE(S)}{1 + sE(S)}$	$s - \lambda [1 - e^{-sE(S)}]$	$\mu s + \frac{1}{2} \sigma^2 s^2$
Solution technique for $\hat{\phi}[\omega(s)] = s$	Newton-Raphson Search	Newton-Raphson Search	Quadratic Formula	Newton-Raphson Search	Quadratic Formula

TABLE 2.3. Summary of Characteristics for the Processes under Study.

model for Brownian motion. Gaver [1968] proposed that this diffusion process could be used as an approximation for the net input process of an M/G/1 queue by equating the first and second moments of the two processes.

The Brownian motion process is denoted by

$$X_B(t) = \sigma \xi(t) - \mu t ,$$

where $\xi(t)$ is a Wiener process whose stationary, independent increments have a normal distribution with mean zero and unit variance. It follows that $X_B(t)$ has mean $-\mu t$ and variance $\sigma^2 t$. As previously noted, the net input process of an M/G/1 queue has mean $E[X(t)] = (\rho-1)t$ and variance $\text{Var}[X(t)] = \lambda E(S^2)t$. Thus, using $X_B(t)$ to approximate $X(t)$ we would take $-\mu = \rho-1$ and $\sigma^2 = \lambda E(S^2)$. The exponent function for Brownian motion is $\Phi(s) = \mu s + \frac{1}{2} \sigma^2 s^2$, cf. Takacs [1967, p. 81]; thus the solution of $\Phi[\omega(s)] = s$ can be calculated using the quadratic formula.

The second non-queueing input process that we consider as an approximation is a gamma input process, denoted

$$X_G(t) = G(t) - t ,$$

where $G(t)$ is a process whose stationary, independent increments have a gamma distribution. The gamma density function is

$$f(x) = \frac{\beta}{\Gamma(\alpha)} (\beta x)^{\alpha-1} e^{-\beta x} , \quad \alpha > 0, \beta > 0, x \geq 0 ,$$

so $G(t)$ has mean $(\alpha/\beta)t$ and variance $(\alpha/\beta^2)t$. It follows directly that $X_G(t)$ has mean $(\alpha/\beta - 1)t$ and variance $(\alpha/\beta^2)t$. Thus, using $X_G(t)$ to approximate the net input process of an M/G/1 queue we would choose α and β such that $\alpha/\beta - 1 = \rho - 1$ and $\alpha/\beta^2 = \lambda E(S^2)$. The exponent function for the gamma input process is $\Phi(s) = s - \alpha \log(1 + s/\beta)$, cf. Takacs [1967, p. 66]; the solution of the functional equation (2.2.3) requires a search technique.

2.3. Scaling the Reflected Levy Processes

In this section we present a method for normalizing the processes of interest by using a simple linear transformation for both the epochs and the server load. This technique allows us to reduce the tabulations required to describe actual operating systems and also facilitates comparisons among different processes by adopting a common, normalized time scale.

Consider an M/G/1 queueing system (as originally described in Section 2.1) with service times S_1, S_2, \dots having distribution function $F(\cdot)$ with mean $E(S)$ and second moment $E(S^2)$. Let $\{A(t); t \geq 0\}$ be the Poisson arrival process with mean arrival rate λ . The net input process is

$$X(t) = S_1 + \dots + S_{A(t)} - t ,$$

and the server load process $W(t)$ is obtained from $[z + X(t)]$ by the reflection mapping described in Section 2.1, where z is the initial server load $W(0)$.

We define a new scaled process in terms of the original queueing process

$$X^*(t) = aX(bt) .$$

This scaled net input process has the form

$$(2.3.1) \quad X^*(t) = S_1^* + \dots + S_{A^*(t)}^* - ct , \quad t \geq 0 ,$$

where $c = ab$, $A^*(t) = A(bt)$, and $S_i^* = aS_i$. Note that the random variables S_i^* have distribution function $G^*(x) = F(x/a)$, and that $A^*(t)$ is a Poisson process with mean arrival rate $\lambda^* = b\lambda$. Let μ^* and σ_*^2 be defined by

$$(2.3.2) \quad E[X^*(t)] = -\mu^* t ,$$

$$\text{Var}[X^*(t)] = \sigma_*^2 t .$$

We accomplish our normalization by choosing a and b so that

$$|\mu^*| = 1 \quad \text{and} \quad \sigma_*^2 = 1 .$$

By substituting $aX(bt)$ for $X^*(t)$ on the left hand side of equations (2.3.2) and then using the parameters of process X as

defined by equations (2.2.1), it is easy to show that this particular scaling requires that

$$a = \frac{|\mu|}{\sigma^2} \quad \text{and} \quad b = \frac{\sigma^2}{\mu^2} .$$

If we define W^* as the reflection of $[z^* + X^*(t)]$, where the scaled initial server load is $z^* = az$, then it is easy to verify that $W^*(t) = aW(bt)$. That is, the reflection of the scaled net input process is identical to the scaled version of the original server load process. If we let t^* represent a (scaled) epoch for the scaled process, then it follows that

$$E_{z^*}[W^*(t^*)] = \frac{|\mu|}{\sigma^2} E_z[W(\frac{\sigma^2}{\mu^2} t^*)] , \quad z = \frac{\sigma^2}{|\mu|} z^* .$$

Thus, we can obtain scaled mean server load by evaluating an original queueing system and applying the transformations.

Before discussing the scaled process further, consider a second queueing system having Poisson arrival process $A'(t)$ with mean rate $\lambda' \neq \lambda$ and service times S'_1, S'_2, \dots with distribution function $F'(x) = F(\lambda'x/\lambda)$ so that $E(S') = \lambda E(S)/\lambda'$. The traffic intensity parameter for this second queueing system is the same as the original, that is,

$$\rho' = \lambda' E(S') = \lambda' \cdot \lambda E(S)/\lambda' = \lambda E(S) = \rho ,$$

defined by equations (2.2.1), it is easy to show that this particular scaling requires that

$$a = \frac{|\mu|}{\sigma^2} \quad \text{and} \quad b = \frac{\sigma^2}{\mu^2} .$$

If we define W^* as the reflection of $[z^* + X^*(t)]$, where the scaled initial server load is $z^* = az$, then it is easy to verify that $W^*(t) = aW(bt)$. That is, the reflection of the scaled net input process is identical to the scaled version of the original server load process. If we let t^* represent a (scaled) epoch for the scaled process, then it follows that

$$E_{z^*}[W^*(t^*)] = \frac{|\mu|}{\sigma^2} E_z[W(\frac{\sigma^2}{\mu^2} t^*)] , \quad z = \frac{\sigma^2}{|\mu|} z^* .$$

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$$\rho' = \lambda' E(S') = \lambda' \cdot \lambda E(S)/\lambda' = \lambda E(S) = \rho ,$$

but the variance parameter is different:

$$\sigma'^2 = \lambda' E(S'^2) = \lambda' \left(\frac{\lambda}{\lambda'}\right)^2 E(S^2) = \frac{\lambda}{\lambda'} \lambda E(S^2) = \frac{\lambda}{\lambda'} \sigma^2 .$$

This second system and the original are related by

$$X'(t) = \frac{\lambda}{\lambda'} X\left(\frac{\lambda'}{\lambda} t\right) ,$$

or equivalently by

$$X'(t) = \frac{E(S')}{E(S)} X\left(\frac{E(S)}{E(S')} t\right) .$$

In the context of $M/E_k/1$ queueing systems, the second queueing system has the same ρ and same Erlang shape parameter k as the original, but the different values for σ^2 and σ'^2 require simple transformation of the epoch scale and waiting time scale.

Returning to the scaled process which was defined in terms of the original queueing process, we now express the original process in terms of the second process:

$$X^*(t) = \frac{|\mu|}{\sigma^2} X\left(\frac{\sigma^2}{\mu^2} t\right) = \frac{|\mu|}{\sigma^2} \frac{\lambda}{\lambda'} X'\left(\frac{\lambda'}{\lambda} \frac{\sigma^2}{\mu^2} t\right) = \frac{|\mu|}{\sigma'^2} X'\left(\frac{\sigma'^2}{\mu^2} t\right) .$$

Once ρ (or μ) has been specified and a specific form of the service time distribution (i.e., a specific Erlang shape parameter k) has been determined, then we are free to choose the variance parameter (i.e.,

the time scales) for the queueing process from which X^* will be obtained.

This research tabulates $E_{z^*}[W^*(t^*)]$ for various values of ρ , k , and z^* , and one can obtain mean server load for an actual queueing system by selecting the table with the closest (ρ, k, z^*) combination or by interpolating between two tabulated scaled processes, and then applying the transformation:

$$E_z[W(t)] = \frac{\sigma^2}{|\mu|} E_{z^*}[W^*(\frac{\mu^2}{\sigma^2} t)] , \quad z^* = \frac{|\mu|}{\sigma^2} z .$$

Since $\mu = 1 - \rho$, this particular normalization cannot be used for systems with $\rho = 1$, i.e., $\mu = 0$. In this case we tabulate mean server load for Erlang queueing systems with λ and $E(S)$ chosen so that the variance parameter $\sigma^2 = \lambda E(S^2) = 1$. Let superscript zero indicate parameters for the tabulated system. Then $\sigma^2 = 1$ requires that

$$\lambda^0 = \frac{k+1}{k} \quad \text{and} \quad E(S^0) = \frac{k}{k+1} .$$

Mean server load for an actual operating system with shape parameter k and mean arrival rate λ can be obtained from the tabulated values as follows:

$$E_z[W(t)] = \frac{\lambda^0}{\lambda} E_{z^0}[W^0(\frac{\lambda}{\lambda^0} t)] = \frac{k+1}{k\lambda} E_{z^0}[W^0(\frac{k\lambda}{k+1} t)] , \quad z^0 = \frac{k\lambda}{k+1} z .$$

The tabulated values for the $\rho = 1$ case correspond to an actual queueing system, whereas the tabulated values for the $\rho \neq 1$ cases represent a stochastic process with "potential work output rate" of $c = ab = 1/|\mu|$, interpreted from equation (2.3.1). For the $\rho < 1$ cases, mean server load in queueing systems approaches the Pollaczek-Khintchine values, equation (2.1.3). Thus, the tabulated values for all scaled processes approach

$$\frac{|\mu|}{\sigma^2} \cdot \frac{\sigma^2}{2|\mu|} = \frac{1}{2} ,$$

thereby allowing comparisons concerning approach to equilibrium.

CHAPTER 3

NUMERICAL INVERSION OF THE LAPLACE TRANSFORM

3.1. Approximation by an Expected Value

In the previous chapter we cited an expression for the Laplace transform for the three processes of interest. For the two special cases of Brownian motion and the M/M/1 queue, analytic methods can be applied and exact numerical solutions can be obtained. However, in general it is necessary to use a method for numerical inversion of the Laplace transform in order to evaluate the process of interest. In this chapter we describe such a method and its implementation.

The first two sections of the chapter describe the method for approximate transform inversion using an expected value. In the third section we present numerical results for test cases where this method is applied to Laplace transforms whose exact inverses are known, and we describe the computer routines that apply the inversion method to prepare tables of $E[W(t)]$.

The method used in this research for numerical inversion of the Laplace transform was originally developed by Gaver [1966]. Although there are other techniques which could have been implemented, this particular method was chosen because it had been applied successfully to the Laplace transform expression for $E[W(t)]$ in the M/M/1 and M/G/1 queues by Gaver [1966, 1968].

The problem of inverting the Laplace transform can be summarized as follows. We have a function of interest, $P(t)$, for which no closed-form analytic expression is known, so that it cannot be evaluated

numerically. However, we do know the expression for its Laplace transform,

$$p(s) = \int_0^{\infty} e^{-st} P(t) dt .$$

Specifically, in this research the function or process of interest is $E[W(t)]$ and the Laplace transform expression was cited in Chapter 2. In general, we wish to tabulate $P(t)$ for various values of t .

The method presented here computes an approximate value of the function at a specified point t' . Theoretically this approximation can be computed as precisely as desired. Consider observing the function of interest at a random time T that has density function $f(t)$ such that the values of T are concentrated near t' , as shown in Figure 3.1. Then $\bar{P}(t')$, an approximation of $P(t')$, can be expressed as

$$(3.1.1) \quad \bar{P}(t') = \int_0^{\infty} P(t) f(t) dt .$$

Now the problem is one of finding an observational density function $f(t)$ so that the right hand side of (3.1.1) can be expressed in terms of $p(s)$.

Gaver [1966] examined such a family of density functions that are well-suited to numerical computation. Let $f_n(t; a)$ be the density function for random variable T , with parameters n and a , as follows:

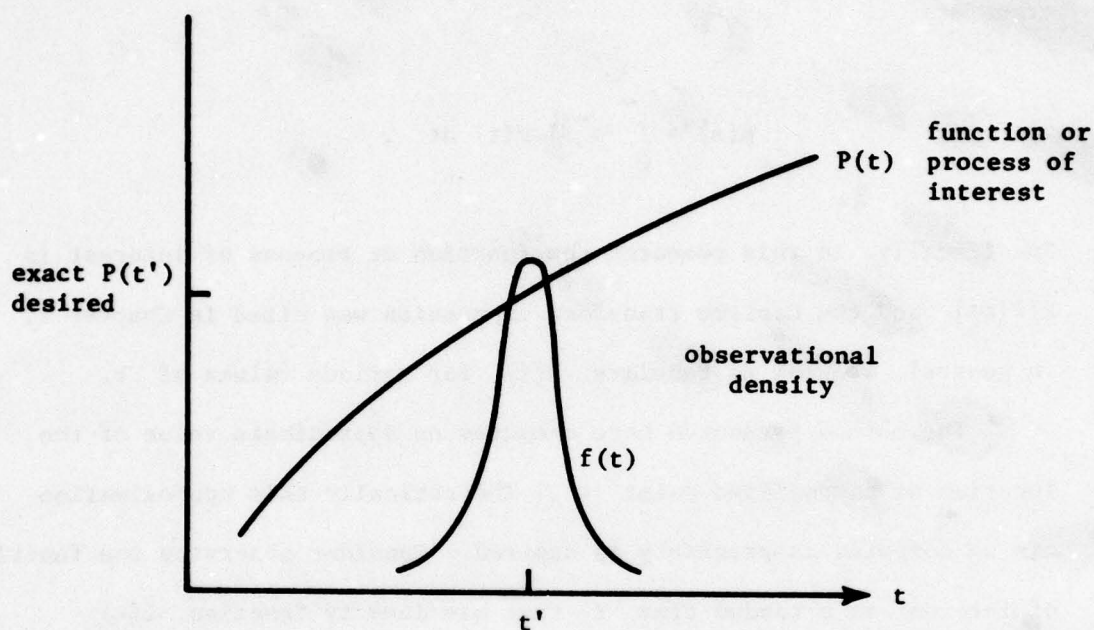


FIGURE 3.1. Observational Density Function

$$(3.1.2) \quad f_n(t; a) = a \frac{(2n)!}{n! (n-1)!} (1 - e^{-at})^n e^{-nat}, \quad a > 0, n = 1, 2, \dots$$

Gaver [1966] showed that $f_n(t; a)$ has the following properties:

$$\text{modal value} \approx \frac{1}{a} \ln 2, \quad ,$$

$$\text{Var}(T) \approx \frac{1}{a^2} \frac{2n+2}{(2n-1)(4n+1)}.$$

For increasing values of n , $f_n(t; a)$ becomes more sharply peaked at $t = 1/a \ln 2$. Using $f_n(t; a)$ as the observational density function

in (3.1.1), we let the approximation of $P(t')$ be

$$(3.1.3) \quad \bar{P}_n = \int_0^{\infty} P(t) f_n(t; a) dt ,$$

where parameter a is chosen such that $a = 1/t' \ln 2$, thereby concentrating the values of random variable T near t' as desired.

By substituting the observational density (3.1.2) in the approximating expression (3.1.3) and then expanding $(1 - e^{-at})^n$ by the binomial theorem, we obtain

$$(3.1.4) \quad \bar{P}_n = a \frac{(2n)!}{n! (n-1)!} \sum_{i=0}^n \binom{n}{i} (-1)^i p[(n+i)a] .$$

Thus, \bar{P}_n , an approximation of $P(t')$, is based on a linear combination of the Laplace transform $p(s)$ evaluated at $n+1$ different values of s . Theoretically, the sequence $\{\bar{P}_n; n = 1, 2, 3, \dots\}$ converges to $P(1/a \ln 2)$, i.e., to $P(t')$. For any finite n , we can calculate \bar{P}_n using (3.1.4), and because $\text{Var}(T) \rightarrow 0$ as $n \rightarrow \infty$ we can obtain as good an approximation as desired by using a large enough value of n . Practically, $p(s)$, the Laplace transform of $P(t)$, must be evaluated at $n+1$ values of s in order to evaluate \bar{P}_n ; in some applications much time may be required to make each evaluation of the Laplace transform expression and so the cost of obtaining the desired precision in \bar{P}_n may be prohibitive. Also, for large n the factorial terms in (3.1.4) are integers with too many digits to be precisely evaluated on computers with finite word length, and operations involving these large numbers

can result in considerable rounding errors. From the standpoint of computational efficiency it is important to improve the accuracy of the approximate inverse without requiring evaluation of \bar{P}_n for extremely large values of n .

3.2. Improving the Accuracy of the Approximation

Gaver [1966] showed that each \bar{P}_n can be represented as an asymptotic expansion, i.e.,

$$(3.2.1) \quad \bar{P}_n \approx P(t') + \frac{\alpha_1}{n} + \frac{\alpha_2}{n^2} + \frac{\alpha_3}{n^3} + \dots$$

where the error components α_i depend only on t' and not on n . An improved approximation of $P(t')$ can be obtained by taking a linear combination of a set of the \bar{P}_n approximations, where the values of n are integer powers of 2. This method, extrapolation to the limit, ensures that some of the error terms in (3.2.1) are cancelled out, cf. Gaver [1966] and Henrici [1964].

Stehfest [1970] demonstrated an improved calculation method that uses a linear combination of an even number of the \bar{P}_n approximations; this method cancels even more of the error terms than Gaver's method (extrapolation to the limit). Stehfest combined the coefficients for the linear combination of \bar{P}_n with the coefficients of $p(s)$ in (3.1.4) so that F_a , the approximation of $P(t')$, could be expressed directly as a linear combination of $p(s)$:

$$(3.2.2) \quad Fa = \frac{\ln 2}{t'} \sum_{i=1}^N v_i p\left(\frac{\ln 2}{t'} i\right) .$$

Here N must be even and the combined coefficients v_i are

$$(3.2.3) \quad v_i = (-1)^{(N/2)+i} \sum_{k=\lceil \frac{i+1}{2} \rceil}^{\min(i, \frac{N}{2})} \frac{k^{N/2} (2k)!}{(\frac{N}{2} - k)! k! (k-1)! (i-k)! (2k-1)!} .$$

For a theoretical comparison of the original \bar{P}_n , extrapolation to the limit, and Fa , consider the three methods where the number of values of $p(s)$ is 32 in each case. Thirty-two evaluations of the $p(s)$ could be used to compute \bar{P}_{31} , and from (3.2.1) the first error term would be $\alpha_{31}/31$. On the other hand the thirty-two values of $p(s)$ could be used to compute $\bar{P}_1, \bar{P}_2, \bar{P}_4, \bar{P}_8$, and \bar{P}_{16} , and it could be shown that extrapolation to the limit would yield an approximation of $P(t')$ where the first four error terms of equation (3.2.1) would cancel completely. The resulting approximation would be $P(t') + \alpha_5/2^{10} + \dots$. Finally using Stehfest's method with $N = 32$, the first fifteen error terms in equation (3.2.1) would cancel, and it could be shown that the approximation would be $Fa = P(t') - \alpha_{16}/16! + \dots$. This theoretical superiority of Stehfest's method was verified numerically in preliminary investigations using functions with known inverses.

3.3. Computer Implementation of the Technique

In this research, the algorithm LINV developed by Stehfest [1970] was translated from ALGOL to FORTRAN IV. Some changes were made in evaluating the V_i of equation (3.2.3) to take advantage of cancelling the factorial terms and thereby avoid rounding errors. Some preliminary numerical investigations used BASIC and single and double precision FORTRAN IV. The computer programs listed in Appendix B specify double precision for all non-integer variables, and the "AUTODBL" option available with the IBM FORTRAN H-Extended Compiler was used in all of the research reported here. With this extended precision there are approximately 36 significant decimal digits for the non-integer variables.

Stehfest [1970] published a table of results applying LINV to six transforms whose inverses are known, using 8 digit arithmetic and $n = 10$. Figure 3.2 shows the exact values of the six functions, the approximations from this research using $N = 34$ with 36 digit arithmetic, and the original Stehfest approximations using $N = 10$. We observe that there are at least six correct significant digits in our approximations (using $N = 34$) for each of the test functions.

Since the LINV approximation (F_a) is based upon the original \bar{P}_n , it should also be more accurate when more values of the Laplace transform $p(s)$ are used, i.e., if N is very large. However, for large values of N the rounding errors in evaluating V_i can affect the accuracy of the results. This problem was investigated by applying LINV to the six test functions of Figure 3.2; approximations were obtained using

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Since the LINV approximation (Fa) is based upon the original \bar{P}_n , it should also be more accurate when more values of the Laplace transform $p(s)$ are used, i.e., if N is very large. However, for large values of N the rounding errors in evaluating V_i can affect the accuracy of the results. This problem was investigated by applying LINV to the six test functions of Figure 3.2; approximations were obtained using

FIGURE 3.2 INVERSION OF TEST FUNCTIONS

T	APPROXIMATE F(T) USING STEHFEST'S METHOD		APPROXIMATE F(T) USING STEHFEST'S METHOD	
	EXACT F(T)	N=34	EXACT F(T)	N=10
1.0	0.5641895543	0.5641895835	-0.5772156715	-0.5772156649
2.0	0.3989422597	0.3989422804	-1.2703628521	-1.2703628455
3.0	0.3257349910	0.3257350079	-1.6758279602	-1.6758279536
4.0	0.2820947771	0.2820947918	-1.9635100327	-1.9635100260
5.0	0.2523132391	0.2523132522	-2.1866535840	-2.1866535773
6.0	0.2303294210	0.2303294330	-2.3689751408	-2.3689751341
7.0	0.2132436076	0.2132436186	-2.5231258206	-2.5231258140
8.0	0.1994711299	0.1994711402	-2.6566572132	-2.6566572066
9.0	0.1880631848	0.1880631943	-2.7744402489	-2.7744402422
10.0	0.1784124024	0.1784124116	-2.8798007645	-2.8798007579
$F(T) = 1/\sqrt{SQRT(PI*T)}$				
1.0	0.5641895543	0.5641895835	-0.5772156715	-0.5772156649
2.0	0.3989422597	0.3989422804	-1.2703628521	-1.2703628455
3.0	0.3257349910	0.3257350079	-1.6758279602	-1.6758279536
4.0	0.2820947771	0.2820947918	-1.9635100327	-1.9635100260
5.0	0.2523132391	0.2523132522	-2.1866535840	-2.1866535773
6.0	0.2303294210	0.2303294330	-2.3689751408	-2.3689751341
7.0	0.2132436076	0.2132436186	-2.5231258206	-2.5231258140
8.0	0.1994711299	0.1994711402	-2.6566572132	-2.6566572066
9.0	0.1880631848	0.1880631943	-2.7744402489	-2.7744402422
10.0	0.1784124024	0.1784124116	-2.8798007645	-2.8798007579
$F(T) = (T*T*T)/6$				
1.0	0.1666666667	0.1666666667	0.3678794412	0.3678794412
2.0	1.3333333333	1.3333333333	0.1353352832	0.1353352832
3.0	4.5000000000	4.5000000000	0.0497870684	0.0497870684
4.0	10.6666666667	10.6666666667	0.0183156389	0.0183156389
5.0	20.8333333333	20.8333333333	0.0067379470	0.0067379470
6.0	36.0000000000	36.0000000000	0.0024787522	0.0024787522
7.0	57.1666666667	57.1666666667	0.0009118820	0.0009118820
8.0	85.3333333333	85.3333333333	0.0003354625	0.0003354625
9.0	121.5000000000	121.5000000000	0.0001234097	0.0001234097
10.0	166.6666666667	166.6666666667	0.0000453999	0.0000453999
$F(T) = \sin(\sqrt{T})$				
1.0	0.9877659460	0.9877659972	-0.6666666667	-0.6666666667
2.0	0.9092974268	0.9092974740	-0.3333333333	-0.3333333333
3.0	0.6381576351	0.6381576682	1.0000000000	1.0000000000
4.0	0.3080717424	0.3080717583	2.3333333333	2.3333333333
5.0	-0.0206835315	-0.0206835326	2.6666666667	2.6666666667
6.0	-0.3169471632	-0.3169471796	1.0000000000	1.0000000000
7.0	-0.5646959281	-0.5646959281	-3.6666666667	-3.6666666667
8.0	-0.7568024953	-0.7568025346	-12.3333333333	-12.3333333333
9.0	-0.8916822545	-0.8916823007	-26.0000000000	-26.0000000000
10.0	-0.9712777590	-0.9712778493	-45.6666666667	-45.6666666667
$F(T) = 1-3*T+3*T^2-T^3$				
1.0	0.9877659460	0.9877659972	-0.6666666667	-0.6666666667
2.0	0.9092974268	0.9092974740	-0.3333333333	-0.3333333333
3.0	0.6381576351	0.6381576682	1.0000000000	1.0000000000
4.0	0.3080717424	0.3080717583	2.3333333333	2.3333333333
5.0	-0.0206835315	-0.0206835326	2.6666666667	2.6666666667
6.0	-0.3169471632	-0.3169471796	1.0000000000	1.0000000000
7.0	-0.5646959281	-0.5646959281	-3.6666666667	-3.6666666667
8.0	-0.7568024953	-0.7568025346	-12.3333333333	-12.3333333333
9.0	-0.8916822545	-0.8916823007	-26.0000000000	-26.0000000000
10.0	-0.9712777590	-0.9712778493	-45.6666666667	-45.6666666667

N equal to 30, 32, 34, and 36. In each case the correct number of decimal places was recorded, rounding off both the exact value and the approximation when making the comparison. Figure 3.3 shows the average number of correct decimal places over the ten values of t evaluated in each case. We observe that the approximations using $N = 34$ are at least as good as those using $N = 32$ or $N = 36$ in all cases. Thus, in the remainder of this research we will use $N = 34$ in the LINV algorithm.¹

Figure 3.4 compares the LINV algorithm with other inversion techniques. The top section of this table reproduces some results by Dubner and by Veillon (1974)²; Dubner's approximations are based on 500 values of $p(s)$, while Veillon's method uses 64 values. We observe that in all cases the LINV algorithm using 34 values of $p(s)$ is at least as accurate as their techniques. The bottom section of Figure 3.4 compares approximations of a second function; we observe that once again the LINV results are as accurate as the results obtained by other techniques.

Figures 3.2, 3.3 and 3.4 indicate that LINV can provide very accurate approximations for the test functions, but it is also important to investigate LINV's accuracy using the function of interest in this

¹In preliminary investigations using FORTRAN IV double precision (approximately 17 digit arithmetic) without the AUTODBL option, the most accurate approximations were obtained using $N = 18$.

²Reprinting privileges for the Stehfest results (Comm. of the ACM, Vol. 13, No. 1, Copyright 1970) and the results published by Veillon (Comm. of the ACM, Vol. 17, No. 10, Copyright 1974) were granted by permission of the Association for Computing Machinery.

FIGURE 3.3. Average Number of Correct Rounded-off Decimal Places in LINV Approximations of Six Functions for Determination of Optimal Value of N

N, the number of values of the Laplace transform $p(s)$ used by LINV to estimate the function

Function	N = 30	N = 32	N = 34	N = 36
$\frac{1}{\sqrt{\pi t}}$	6.9	6.9	6.9	6.9
$\frac{t^3}{6}$	9.7	10.5	12.7	11.9
$\sin \sqrt{2t}$	6.7	6.7	6.7	6.7
$-C - \ln t$	7.3	7.3	7.5	7.4
e^{-t}	9.7	10.1	10.5	10.0
$1 - 3t + \frac{3t^2}{2} + \frac{t^3}{6}$	9.7	10.5	11.9	10.6

FIGURE 3.4 COMPARISONS WITH OTHER TECHNIQUES

$$F(T) = 1/\sqrt{T(T+1)}$$

APPROXIMATE F(T) USING NUMERICAL INVERSION	
STEHFEST'S METHOD	
T	VEILLON*
1	0.56419
2	0.39894
3	0.32573
4	0.28209
5	0.25231
6	0.23033
7	0.21324
8	0.19947
9	0.18806
10	0.17841

$$F(T) = \exp(-T/2)$$

APPROXIMATE F(T) USING NUMERICAL INVERSION	
STEHFEST N=34	
T	VEILLON*
4.140186	0.126174
2.501126	0.286329
1.643438	0.439675
1.085084	0.581269
0.693147	0.707107
0.412298	0.813712
0.214821	0.898158
0.085541	0.958135
0.016048	0.992015

*NOTE: DATA FOR METHODS MARKED BY AN ASTERISK WERE TAKEN FROM A.C.M. ALGORITHM 486. *NUMERICAL INVERSION OF LAPLACE TRANSFORM. BY FRANCOISE VEILLON, COMMUNICATIONS OF THE A.C.M., VOLUME 17, NUMBER 10, OCTOBER 1974.

research, $E[W(t)]$. Our earlier investigations using LINV (not included here) verified the results for M/M/1 queues reported by Gaver [1968], but it is more appropriate to compare our LINV results with mean server load obtained by some method other than Laplace transform inversion. Fortunately, Coleman [1975] evaluated $E[W(t)]$ in the M/M/1 queue using a sum of Bessel functions. In Figure 3.5 the first three columns show his results at five epochs and the LINV results using equations (2.2.4) and (2.2.5) for the Laplace transform expression. The five significant digits available in the Coleman results are matched exactly when the LINV results are rounded off.

As mentioned in Section 2.2, except for the Wiener process and the M/M/1 queue, the value of $\omega(s)$ must be obtained using a search technique. Our final results use the quadratic formula to determine $\omega(s)$ for the Wiener process, but in all other cases (M/D/1, M/E_k/1 including M/M/1, and the gamma input process) we use the standard Newton-Raphson method. This search technique was chosen because expressions for the derivatives of the functions were available and subsequent computations demonstrated that the search usually converged in only four or five iterations.

The exponent function $\phi(\cdot)$ is designated FMD1, FMEK, or FGAM in the FORTRAN subroutines, and functions FOMD1, FOMEK, and FOGAM are defined as

$$FO \cdots [\omega(s)] = F \cdots [\omega(s)] - s \quad .$$

Thus, in order to determine $\omega(s) > 0$ such that $\Phi[\omega(s)] = s$, we search for the value of ω such that $F_0 \cdots$ equals zero. Subroutine ZERO performs the search, computing the relative accuracy (change in ω from the last iteration). When the absolute value of the relative accuracy is less than a specified tolerance, the search stops, and the current value of ω is used to compute the Laplace transform using equation (2.2.4).

The choice of tolerance value (FORTRAN variable TOLRNC) affects the accuracy of ω determined from the search, and the accuracy of ω affects $p(s)$ and the estimate of $E[W(t)]$. Figure 3.5 shows both the approximations of $E[W(t)]$ based upon the exact quadratic solution of the functional equation (2.2.4) and the approximations obtained using the Newton-Raphson search with various specified tolerances. We observe that the two results agree for nine or ten significant digits in all cases and that the number of search iterations does not increase greatly as the tolerance is decreased. For each of the ten epochs, $\omega(s)$ is evaluated thirty-four times ($N = 34$); thus, the total number of iterations shown in Figure 3.5 applies to 340 evaluations of $\omega(s)$. Since the average number of iterations in each evaluation only increases from 4.2 to 5.3 as the tolerance decreases from 10^{-10} to 10^{-24} , we have set the search tolerance at 10^{-20} .

In Chapter 2 we discussed the stochastic process $W(t)$ in an M/G/1 queue and its expected value function. Equation (2.1.3) is restated here in a slightly different form along with the Laplace transform of $E_z[W(t)]$, equation (2.2.4):

Thus, in order to determine $\omega(s) > 0$ such that $\Phi[\omega(s)] = s$, we search for the value of ω such that $F_0 \cdots$ equals zero. Subroutine ZERO performs the search, computing the relative accuracy (change in ω from the last iteration). When the absolute value of the relative accuracy is less than a specified tolerance, the search stops, and the current value of ω is used to compute the Laplace transform using equation (2.2.4).

The choice of tolerance value (FORTRAN variable TOLRNC) affects the accuracy of ω determined from the search, and the accuracy of ω affects $p(s)$ and the estimate of $E[W(t)]$. Figure 3.5 shows both the approximations of $E[W(t)]$ based upon the exact quadratic solution of the functional equation (2.2.4) and the approximations obtained using the Newton-Raphson search with various specified tolerances. We observe that the two results agree for nine or ten significant digits in all cases and that the number of search iterations does not increase greatly as the tolerance is decreased. For each of the ten epochs, $\omega(s)$ is evaluated thirty-four times ($N = 34$); thus, the total number of iterations shown in Figure 3.5 applies to 340 evaluations of $\omega(s)$. Since the average number of iterations in each evaluation only increases from 4.2 to 5.3 as the tolerance decreases from 10^{-10} to 10^{-24} , we have set the search tolerance at 10^{-20} .

In Chapter 2 we discussed the stochastic process $W(t)$ in an M/G/1 queue and its expected value function. Equation (2.1.3) is restated here in a slightly different form along with the Laplace transform of $E_z[W(t)]$, equation (2.2.4):

FIGURE 3.5 EFFECT OF NEWTON-RAPHSON SEARCH TOLERANCE

THIS TABLE COMPARES CCLEMAN'S BESSEL FUNCTION RESULTS AT FIVE EPOCHS WITH LAPLACE INVERSION RESULTS FOR BOTH THE EXACT QUADRATIC SOLUTION OF THE FUNCTIONAL AND THE APPROXIMATE NEWTON-RAPHSON SEARCH SOLUTION OF THE FUNCTIONAL WITH VARIOUS TOLERANCES.

M/M/1 MEAN SERVER LOAC. LAMEDA = 1.0. E(S) = 0.95. Z = 0.

EPOCH T	SUMS OF BESSEL FUNCTIONS (COLEMAN)	INVERSION OF LAPLACE TRANSFORM, STEHFEST'S METHOD, N=34			
		QUADRATIC SOLUTION CF FUNCTIONAL		TOLERANCE FOR NEWTON-RAPHSON SEARCH SOLUTION OF FUNCTIONAL	
		10**-10	10**-12	10**-14	10**-16
20.	3.9181	3.91805123011	3.91805123011	3.91805123011	3.91805123011
40.	5.4651	5.46511190320	5.46511190320	5.46511190320	5.46511190320
60.	6.5582	6.55818100638	6.55818100638	6.55818100638	6.55818100638
80.	7.4191	7.41908998154	7.41908998154	7.41908998154	7.41908998154
100.	8.1335	8.13354320450	8.13354320450	8.13354320450	8.13354320450
200.		10.56878873791	10.56878873791	10.56878873791	10.56878873791
300.		12.08196260537	12.08196260537	12.08196260537	12.08196260537
400.		13.15027687639	13.15027687639	13.15027687639	13.15027687639
800.		15.49230830918	15.49230830918	15.49230830917	15.49230830920
1200.		16.56195996067	16.56195996093	16.56195996058	16.56195996061
TOTAL NUMBER OF NEWTON-RAPHSON ITERATIONS TO EVALUATE 10 EPOCHS:		1423	1460	1518	1602

EPOCH T	SUMS OF BESSEL FUNCTIONS (COLEMAN)	INVERSION OF LAPLACE TRANSFORM, STEHFEST'S METHOD, N=34			
		QUADRATIC SOLUTION CF FUNCTIONAL		TOLERANCE FOR NEWTON-RAPHSON SEARCH SOLUTION OF FUNCTIONAL	
		10**-18	10**-20	10**-22	10**-24
20.	3.9181	3.91805123011	3.91805123011	3.91805123011	3.91805123011
40.	5.4651	5.46511190320	5.46511190320	5.46511190320	5.46511190320
60.	6.5582	6.55818100638	6.55818100638	6.55818100638	6.55818100638
80.	7.4191	7.41908998154	7.41908998154	7.41908998154	7.41908998154
100.	8.1335	8.13354320450	8.13354320450	8.13354320450	8.13354320450
200.		10.56878873791	10.56878873791	10.56878873791	10.56878873791
300.		12.08196260537	12.08196260537	12.08196260537	12.08196260537
400.		13.15027687639	13.15027687639	13.15027687638	13.15027687638
800.		15.49230830918	15.49230830918	15.49230830918	15.49230830920
1200.		16.56195996067	16.56195996062	16.56195996061	16.56195996068
TOTAL NUMBER OF NEWTON-RAPHSON ITERATIONS TO EVALUATE 10 EPOCHS:		1716	1756	1773	1794

$$E_z[W(t)] = z - (1-p)t + E_z[I(t)] ,$$

$$P_W(z,s) = \frac{z}{s} - \frac{\mu}{s^2} + \frac{e^{-\omega(s)z}}{s(s)} ,$$

The first two terms of $P_W(z,s)$ can be inverted "by inspection," i.e., by referring to any table of function-transform pairs. Only the last term requires numerical inversion using LINV. Stehfest [1970, p. 48] cautioned the prospective user of LINV as follows: "One ought to be sure a priori that the unknown function $F(t)$ has not any discontinuities, salient points, sharp points, sharp peaks, or rapid oscillations."

Recall that our unknown function, $E_z[I(t)]$ in M/G/1 queues, is a non-decreasing function and that it must be zero between $t = 0$ and $t = z$. Although it is not a discontinuous function, its slope does change abruptly at $t = z$, and we might expect some inaccuracies near that point. The top section of Figure 3.6 shows scaled approximations of $E_z[I(t)]$ evaluated at $t = z$ for M/M/1 queues with traffic intensities between .5 and 2 and scaled initial server loads between .2 and 4. If the inversion technique is accurate, then all entries in the table should be zero. We observe that there are some inaccuracies, particularly for queues with low traffic intensities. For example, the worst case is $\rho = .5$ with $z' = 1.0$, where scaled $E_z[I(t)]$ is .007 instead of zero.

Since the inaccuracies may be related to the discontinuity of the derivative of $E_z[I(t)]$ at $t = z$, our approach does not invert the third term of $P_W(z,s)$ directly. Instead, we construct a new function,

FIGURE 3.6 CORRECTIONS FOR DISCONTINUOUS FIRST DERIVATIVE

APPROXIMATIONS USING STEHFEST'S METHOD. N=34

M/M/1 SCALED MEAN CUMULATIVE IDLENESS AT $T = \text{ABS}(\text{MU}) * Z$ (I.E., AT $T=Z$)

WITHOUT CORRECTION TERM										
SCALED INITIAL SERVER LOAD										
	0.2	0.4	0.6	0.8	1.0	2.0	3.0	4.0		
RHO										
0.5	0.003122	0.005112	0.006274	0.006845	0.007000	0.005125	0.002718	0.001107		
0.6	0.002661	0.003710	0.003879	0.003605	0.003140	0.000965	0.000101	0.000093		
0.7	0.001984	0.002061	0.001606	0.001108	0.000707	0.000034	0.000021	0.000001		
0.8	0.001059	0.000587	0.000235	0.000072	0.000010	0.000001	0.000000	0.000000		
0.9	0.000148	0.000006	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000		
1.0	0.000029	0.000001	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000		
1.1	0.000014	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000		
1.2	0.000007	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000		
1.3	0.000004	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000		
1.4	0.000002	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000		
1.5	0.000001	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000		
2.0	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000		

M/M/1 SCALED MEAN CUMULATIVE IDLENESS AT $T = \text{ABS}(\text{MU}) * Z$ (I.E., AT $T=Z$)

WITH CORRECTION TERM										
SCALED INITIAL SERVER LOAD										
	0.2	0.4	0.6	0.8	1.0	2.0	3.0	4.0		
RHO										
0.5	0.000002	0.000002	0.000001	0.000006	0.000010	0.000033	0.000129	0.000289		
0.6	0.000002	0.000000	0.000004	0.000007	0.000010	0.000076	0.000157	0.000150		
0.7	0.000001	0.000000	0.000004	0.000009	0.000020	0.000060	0.000024	0.000001		
0.8	0.000001	0.000000	0.000001	0.000019	0.000021	0.000001	0.000000	0.000000		
0.9	0.000001	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000		
1.0	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000		
1.1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000		
1.2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000		
1.3	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000		
1.4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000		
1.5	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000		
2.0	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000		

$$H(t) = E_z[I(t)] + g(t) ,$$

where $g(t)$ and its Laplace transform are known, and where $g(t)$ is chosen so that both $H(t)$ and its derivative are continuous. Then LINV is used to invert the Laplace transform of $H(t)$, and approximate $E_z[I(t)]$ is obtained by subtracting $g(t)$ from approximate $H(t)$.

It can be shown that the slope of $E_z[I(t)]$ in M/G/1 queues is equal to the probability that the server load is zero, denoted $P\{W(t) = 0\}$, at any epoch t . Therefore, the slope of $E_z[I(t)]$ at $t = z$ is $P\{W(z) = 0\}$, which is the probability that no arrivals will occur between $t = 0$ and $t = z$, or $e^{-\lambda z}$. For $H(t)$ to have a continuous first derivative at $t = z$, that derivative must be zero. Since the slope of $E_z[I(t)]$ is $e^{-\lambda z}$ at that point, then the slope of $g(t)$ must be $-e^{-\lambda z}$. Therefore, we define $g(t)$ as follows:

$$g(t) = \begin{cases} 0 & \text{for } 0 \leq t \leq z \\ -e^{-\lambda z}(t-z) , & \text{for } t > z \end{cases}$$

Referring to any table of transform-function pairs, the Laplace transform of $g(t)$ is $-\exp[-z(\lambda+s)]/s^2$. Thus the Laplace transform of $H(t)$, i.e., the transform of $E_z[I(t)] + g(t)$, is

$$(3.3.1) \quad P_H(z,s) = \int_0^{\infty} e^{-st} H(t) dt = \frac{e^{-\omega(s)z}}{s\omega(s)} - \frac{e^{-z(\lambda+s)}}{s^2} .$$

Applying LINV to $P_H(z,s)$ we obtain an approximation of $H(t)$. Using overscore to denote approximations, we compute the estimate of $E_z[I(t)]$ as follows:

$$\bar{E}_z[I(t)] = \bar{H}(t) - g(t) = \bar{H}(t) + \begin{cases} 0 & \text{for } 0 \leq t \leq z \\ e^{-\lambda z}(t-z) & \text{for } t > z \end{cases}$$

The bottom section of Figure 3.6 shows scaled $\bar{E}_z[I(t)]$ when the correction term, $g(t)$, is included in the analysis. The results are considerably improved, and since the main tables will express $E_z[W(t)]$ in hundredths or thousandths, the slight inaccuracies which remain will not affect our tabulated results.

The correction term, $g(t)$, is employed in our computations for both $M/D/1$ and $M/E_k/1$ queues. Equation (3.3.1) is incorporated into subprogram functions PMD1 and PMEK of the main programs. The gamma input process does not require the correction term because both $E_z[I(t)]$ and its derivative are continuous, i.e., the slope of $E_z[I(t)]$ is zero at $t = z$. The method is not applied to the Wiener process because process $W(t)$ cannot be separated into four components, $z + S(t) - t + I(t)$; thus, the third term of $P_W(z,s)$ cannot be interpreted as the Laplace transform of expected cumulative idleness in this case.

So far, our discussions in this section have covered some of the important details of the two main computer programs which produce the tables of scaled $E_z[W(t)]$ in Appendix C. Listings of these two main

programs, one for $\rho = 1$ and one for $\rho \neq 1$, are included in Appendix B. We now discuss the algorithm for $M/E_k/1$ queues with $\rho \neq 1$, and we briefly mention the differences in the algorithms for the other three processes and for $\rho = 1$. Figure 3.7 describes both the steps of the algorithm and the names of the subprograms which perform those steps.

The first program specifies the values of the parameters for subroutines LINV and ZERO. Values of ρ , scaled initial server load (z^*), and sixteen scaled epochs (t^*) are read from punched cards. The program performs computations for three pages of tables on each run. The three tables on a page have the same value of ρ and epochs t^* , but each table has a different value of z^* . For each table we compute $\bar{E}_z[W(t)]$ for the Wiener process at all sixteen epochs, then for the $M/D/1$ queue at the same sixteen epochs, followed by each of the seven $M/E_k/1$ queues and the gamma input process.

For each of the two processes the first step is conversion of z^* and t^* to unscaled values. As discussed in Section 2.3, the time scale conversion factor is $(1-\rho)^2/\sigma^2$; this factor is denoted ALFSQR in the subprograms RZWNR, RZMD1, RZMEK, and RZGAM. The value of σ^2 for the unscaled process is selected so that the formulas and numerical computations are subsequently simplified. The conversion factor for server load or wait is $|\mu|/\sigma^2$, denoted WTFCTR in the subprograms.

The initial trial value of $\omega(s)$ in the Newton-Raphson search for the solution of $\Phi[\omega(s)] = s$ is denoted OSTART. For a specific z^* and $\rho \leq 1$, OSTART is arbitrarily set equal to 1 for the first

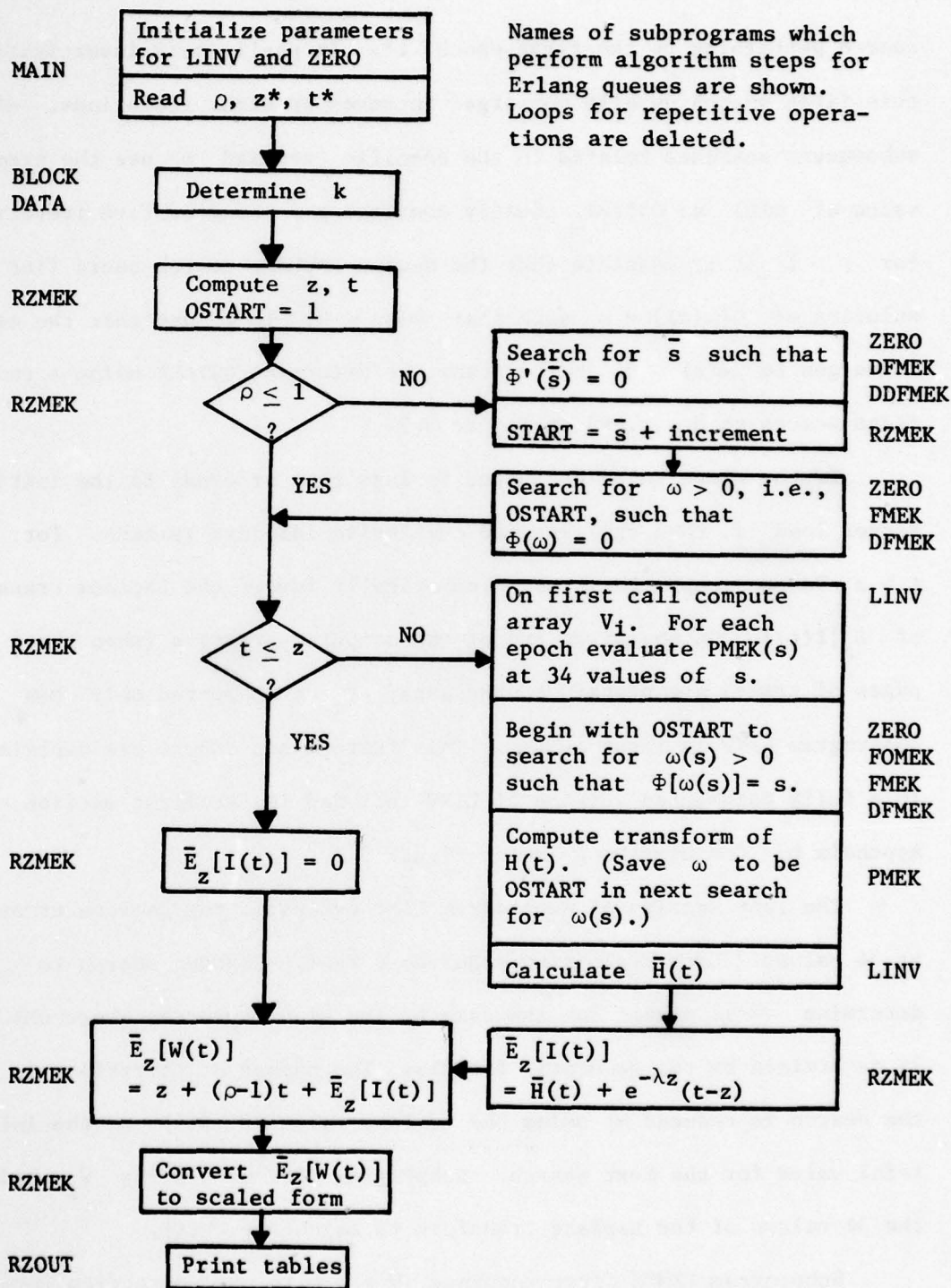


FIGURE 3.7. Flowchart of Algorithm for Tables

search pertaining to the first epoch t^* . In preliminary investigations this first search usually converged in seven or eight iterations. All subsequent searches related to the specific z^* and ρ use the previous value of $\omega(s)$ as OSTART, usually converging in four or five iterations. For $\rho > 1$ it is possible that the Newton-Raphson search could find a solution of $\Phi[\omega(s)] = s$ such that $\omega(s) < 0$. To ensure that the search converges to $\omega(s) > 0$ in this case, we determine OSTART using a two-stage search as described in Figure 3.7.

If the epoch to be evaluated is less than or equal to the initial server load z , then the expected cumulative idleness is zero. For $t > z$ we use subprogram LINV to numerically invert the Laplace transform of $E_z[I(t)]$. On any given run of the computer programs (when three pages of tables are prepared), the array V_i is computed only when subprogram LINV is first called. This feature and others are explained in a fully documented version of LINV included in the first section of Appendix B, "Computer Program for Figure 3.2."

The last section of subprogram LINV evaluates the Laplace transform at 34 values. Each evaluation requires a Newton-Raphson search to determine $\omega(s)$, except for the case of the Wiener process where OMEGA is determined by the quadratic formula. The number of iterations in the search is reduced by using the current value of $\omega(s)$ as the initial trial value for the next search. Subprogram LINV uses array V_i and the 34 values of the Laplace transform to calculate $\bar{H}(t)$.

Subprogram RZMEK first combines $\bar{H}(t)$ with the correction term to obtain $\bar{E}_z[I(t)]$, and then it computes $\bar{E}_z[W(t)]$ for the unscaled

process. In the final step $\bar{E}_z[W(t)]$ is converted to scaled form and stored in the three-dimensional array SCLWT. When the computations for three tables (each with ten processes evaluated at sixteen epochs) are completed, subprogram RZOUT prints five copies of the tables with varying margins. Then the main program reads punched cards specifying ρ , z^* , and t^* for the next page of tables.

All computer programs listed in Appendix B were coded in IBM FORTRAN IV [IBM, 1971] and processed on an IBM 370/168 with the high-speed-multiply feature. The average computer time needed to evaluate scaled $E_z[W(t)]$ for one process at one epoch, i.e., a single value in the tables of Appendix C, was approximately 0.17 second.

CHAPTER 4

NUMERICAL RESULTS FOR EXPECTED SERVER LOAD

4.1. Using the Tables

In this section we explain how the tables in Appendix C can be used to obtain time-dependent mean server load in $M/E_k/1$ queues. We will repeat only the notation and formulas from the previous chapters that are required for using the tables; the details of the underlying theory and numerical method will not be discussed here.

The system being studied is the standard $M/E_k/1$ queue. We assume that the analyst has specified the value of the mean arrival rate, denoted λ , and has determined that the service times (random variable S) can be described by a gamma or Erlang distribution. If an actual operating system is being studied and data are available, the analyst can use the methods described by Hora [1978] or Reinmuth [1971] to verify that the input process is Poisson. Similarly, a chi-square goodness-of-fit test can be used to compare the actual distribution of service times with tabulated values of the incomplete gamma function [Pearson, 1922]. In cases where the service times of an actual system do not exactly fit the gamma distribution, the analyst still may wish to use this theoretical distribution as an approximation.

We define the probability density function for this two-parameter gamma distribution using the mean, $E(S)$, and the Erlang shape parameter, k (not restricted to integer values), as follows:

$$f(t) = \frac{[k/E(S)]^k}{(k-1)!} t^{k-1} e^{-kt/E(S)}, \quad t \geq 0, k > 0.$$

The standard deviation of this service time distribution is $E(S)/\sqrt{k}$, so the appropriate value of parameter k can be determined either from the mean and second moment, $E(S^2)$, or from the mean and variance, $\text{Var}(S)$, as follows:

$$(4.1.1) \quad k = \frac{[E(S)]^2}{\text{Var}(S)} = \frac{[E(S)]^2}{E(S^2) - [E(S)]^2}.$$

Figure 4.1 shows two groups of gamma service time distributions, where the special case of the exponential distribution ($k = 1$) is included in both groups. The mean, $E(S)$, of each distribution in the figure is equal to unity, so the standard deviation is $1/\sqrt{k}$ in each case. The top chart includes density functions with $k = 2$ and $k = 4$, i.e., with less variance than the exponential distribution. The limiting case as k becomes infinitely large is the degenerate distribution with constant service times, corresponding to the M/D/1 queue. The bottom chart of Figure 4.1 includes density functions with $k = .5$ and $k = .2$; these two distributions can be used to approximate systems where the service times have more variability than the exponential case.

The stochastic process being studied is server load, denoted $W(t)$. This process is also called unfinished work, system backlog, or virtual waiting time, the latter because $W(t)$ represents the time that an arrival at epoch t would wait before beginning service. The system may begin operation at epoch $t = 0$ with some initial backlog, $W(0) = z$.

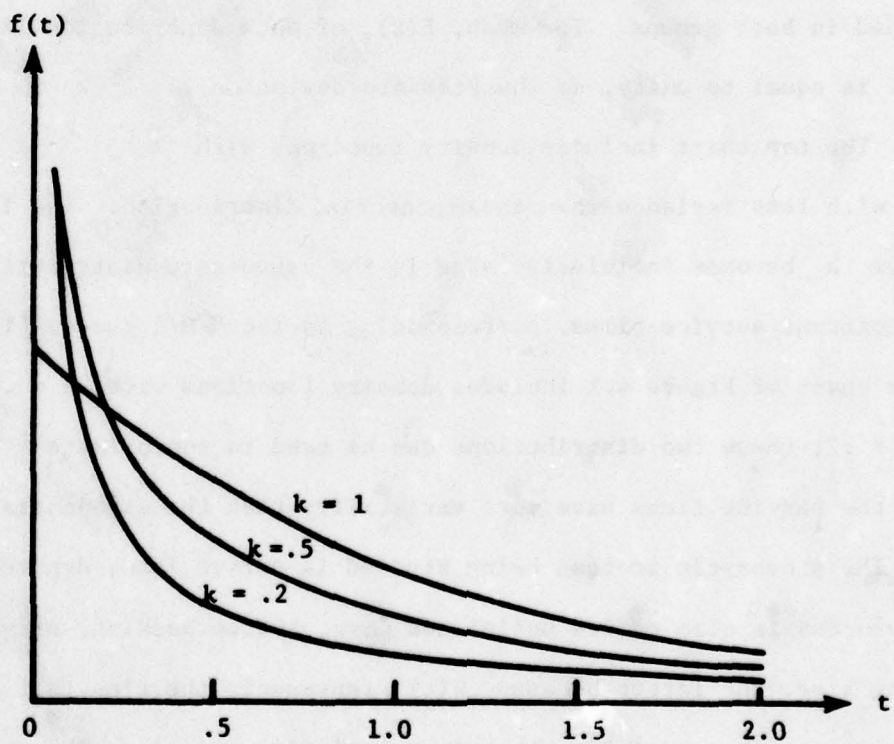
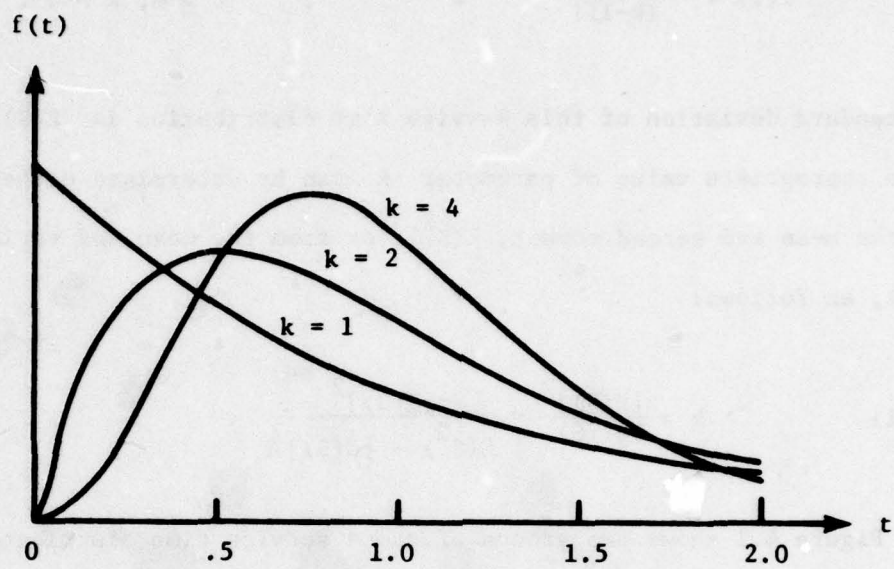


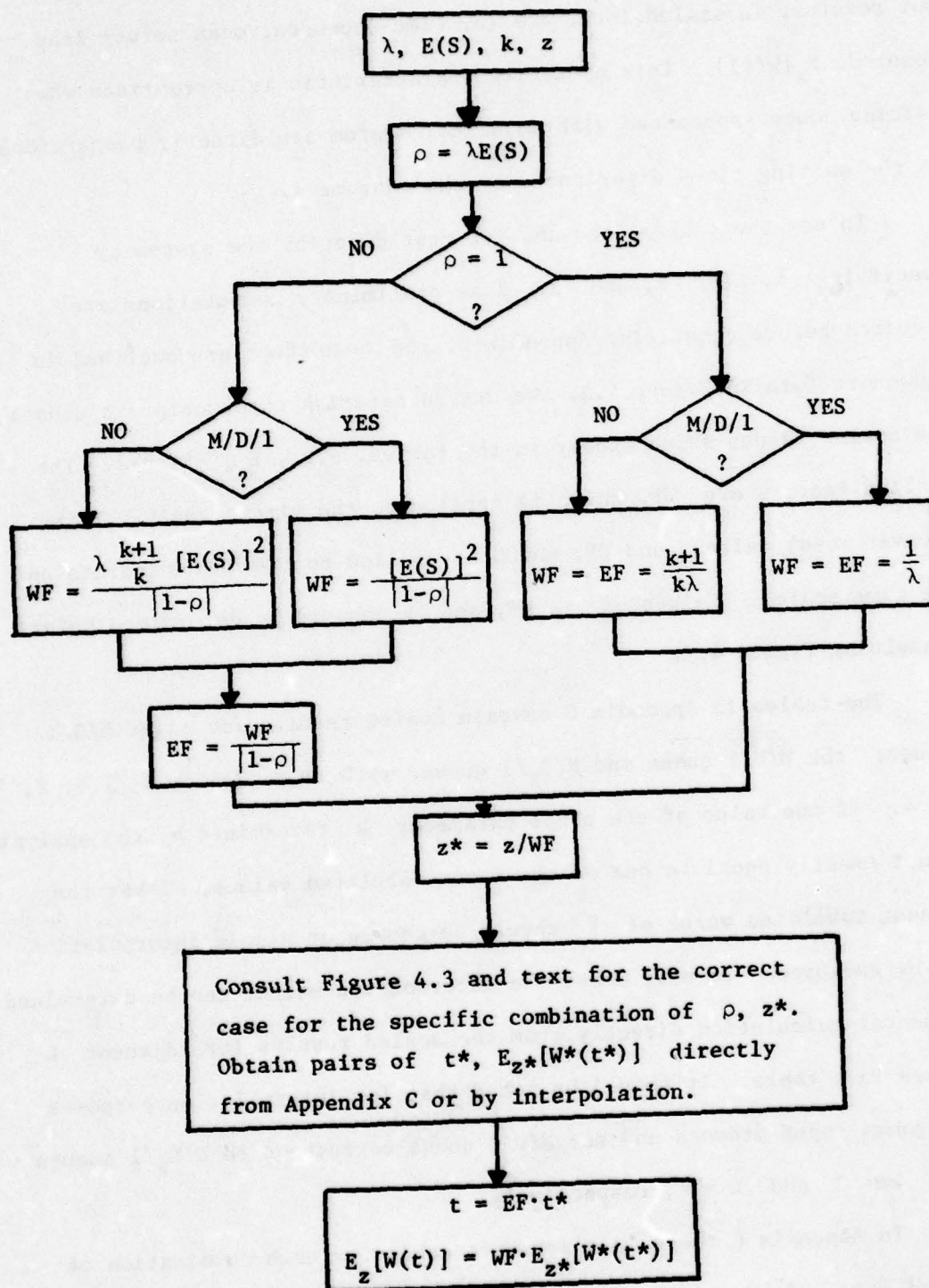
FIGURE 4.1. Erlang (Gamma) Density Functions with Mean = 1.

Our results, in scaled form, are for time-dependent mean server load, denoted $E_z[W(t)]$. This operating characteristic is appropriate when waiting costs associated with an actual system are directly proportional to the waiting times experienced by the customers.

To use the tables, the analyst must describe the system by specifying λ , $E(S)$, k , and z . Some preliminary computations are required before consulting Appendix C, and these steps are outlined in flowchart form in Figure 4.2. We use an asterisk superscript to denote the scaled values which appear in the tables, e.g., $E_{z*}[W^*(t^*)]$. The scaling factors are WF , which is applied to the virtual waiting time (server load) values, and EF , which is applied to the epochs (points on the time scale). Values of ρ , z^* , and k should be determined before consulting Figure 4.3.

The tables in Appendix C contain scaled results for eight M/G/1 queues: the M/D/1 queue and $M/E_k/1$ queues with $k = .2, .5, 1, 1.5, 2, 3$, and 4. If the value of the shape parameter k determined by the analyst is not exactly equal to one of the seven tabulated values, either the closest tabulated value of k should be chosen or simple interpolation can be employed. In most cases the interpolated values can be determined by mental calculation directly from the scaled results for adjacent k values in a table. It should be noted that for interpolation purposes the gamma input process and the M/D/1 queue correspond to $M/E_k/1$ queues with $k = 0$ and $k = \infty$, respectively.

In Appendix C there is a separate table for each combination of ρ and z^* . If the analyst chooses not to use the closest (ρ, z^*)



Note: EF = Epoch Factor for scaling
 WF = Wait Factor for scaling

FIGURE 4.2. Flowchart Instructions for Using the Tables

	$z^* = 0, .2, .4, .6, .8, 1.0$	$0 < z^* < 1; z^* \neq .2, .4, .6, .8$	$z^* = 2., 3., 4.$	$1 < z^* < 4; z^* \neq 2., 3.$	$z^* > 4.$
$\rho < .5$	I	I	I	I	I
$\rho = .5, .6, .7, .8, .9$	A	C	A	C	H
$.5 < \rho < .9; \rho \neq .6, .7, .8$	B	D	B	D	H
$\rho = 1.0$	A	C	A	C	G
$\rho = 1.1, 1.2, 1.3, 1.4, 1.5, 2.$	A	C	F	F	F
$1.1 < \rho < 2.; \rho \neq 1.2, 1.3, 1.4, 1.5$	B	D	F	F	F
$\rho > 2.0$	E	E	F	F	F

FIGURE 4.3. Various Cases for (ρ, z^*) Combinations

combination which is tabulated, then interpolation between two or more tables will be required. Figure 4.3 illustrates the cases which might be encountered, and the appropriate methods for these cases are discussed below. For any specific (ρ, z^*) combination being studied, the objective is to determine pairs of t^* and $E_{z^*}[W^*(t^*)]$ to which the scaling factors, EF and WF, will be applied.

Case A. No interpolation is required. There is a table in Appendix C for this specific (ρ, z^*) combination, and the values for scaled epochs, t^* , and scaled mean server load, $E_{z^*}[W^*(t^*)]$, can be read directly from the table.

Case B. The exact value of z^* is tabulated, but the exact value of ρ falls between two tabulated values. First find the two tables that have the exact z^* with the two closest values of ρ . Then list the scaled epochs (t^*) that appear on both tables, and also list both values of $E_{z^*}[W^*(t^*)]$ for each of the common epochs. For each epoch use simple interpolation to determine $E_{z^*}[W^*(t^*)]$ for the exact value of ρ .

Case C. The exact value of ρ is tabulated, but the exact value of z^* falls between two tabulated values. The method is similar to Case B. Find the two tables with the exact ρ and the two closest values of z^* ; list the common values of t^* with the two values of $E_{z^*}[W^*(t^*)]$; and use simple interpolation. The initial behavior of

the system can also be determined as follows:

$$E_{z^*}[W^*(t^*)] = \begin{cases} z^* - t^* & \text{for } \rho < 1, 0 \leq t^* \leq (1-\rho)z^* \\ z^* & \text{for } \rho = 1, 0 \leq t^* \leq z^* \\ z^* + t^* & \text{for } \rho > 1, 0 \leq t^* \leq (\rho-1)z^* \end{cases}$$

Case D. The exact values of neither ρ nor z^* are tabulated, but the exact value of each does fall between two tabulated values. For one of the closest z^* values which are tabulated, perform the interpolation between the two closest values of ρ using the procedure of Case B. Use the same procedure for the other closest value of z^* . Finally, use the procedure of Case C to interpolate between those two sets of data. For example, if we are interested in results for $(\rho = .67, z^* = .44)$, we interpolate between $(\rho = .6, z^* = .4)$ and $(\rho = .7, z^* = .4)$ to obtain $(\rho = .67, z^* = .4)$. We also interpolate between $(\rho = .6, z^* = .6)$ and $(\rho = .7, z^* = .6)$ to obtain $(\rho = .67, z^* = .6)$. Finally, we interpolate between $(\rho = .67, z^* = .4)$ and $(\rho = .67, z^* = .6)$ to obtain results for $(\rho = .67, z^* = .44)$.

Case E. For situations where $\rho > 2$ and $0 < z^* < 1$, the Wiener process provides an upper bound for $E_{z^*}[W^*(t^*)]$ in any queueing system. Because of the scaling used in the tables, the tabulated values of $E_{z^*}[W^*(t^*)]$ for the Wiener process depend only on z^* , not on ρ . Therefore, any table with $\rho \geq 1.1$ and the specified value of z^* can be used to determine the upper bound. Interpolation can be used if the

exact value of z^* falls between the tabulated values. In these same situations a lower bound for $E_{z^*}[W^*(t^*)]$ is $z^* + t^*$.

Case F. For situations with both $\rho > 1$ and $z^* > 1$, a very accurate approximation of $E_{z^*}[W^*(t^*)]$ is $z^* + t^*$. The accuracy is best when both ρ and z^* are large, but even for $\rho = 1.1$ and $z^* = 1$ the maximum error (M/D/1 queue, $t^* = 1.5$) is only 1.4 percent.

Case G. For $\rho = 1$ and $z^* > 4$, the initial behavior is $E_{z^*}[W^*(t^*)] = z^*$ for $0 \leq t^* \leq z^*$. For $t^* > z^*$, the results for $z^* = 4$ provide a loose lower bound.

Case H. For $.5 \leq \rho \leq .9$ and $z^* > 4$, the initial behavior is $E_{z^*}[W^*(t^*)] = z^* - t^*$, for $0 \leq t^* \leq (1-\rho)z^*$. For $t^* > (1-\rho)z^*$, the results for $z^* = 4$ provide a lower bound.

Case I. For systems with $\rho < .5$ the approach to equilibrium is very slow. If the initial server load $z^* > 0$, the initial behavior is the same as Case H. For $z^* \leq 4$ the results for the same queue (i.e., the same value of k) with $\rho = .5$ provide an upper bound.

Other Cases. For situations with $.9 < \rho < 1.0$ and $0 \leq z^* \leq 4$, the tabulated results for the Wiener process with any $\rho < 1$ and the appropriate z^* provide a tight upper bound for any queueing system. Likewise, for situations with $1.0 < \rho < 1.1$ and $0 \leq z^* \leq 4$, the

tabulated results for the Wiener process with any $\rho > 1$ and the appropriate z^* provide a tight upper bound. In both cases the Wiener approximation is most accurate for values of ρ closest to unity.

The tables in Appendix C can be used along with Figures 4.2 and 4.3 to obtain explicit results for the transient behavior of mean server load in a wide range of M/G/1 queues. In many applications the analyst will perform some sensitivity analysis by using various estimates of λ , $E(S)$, k , and z and comparing the results. In cases where $\rho < 1$ the analyst may be interested only in the speed with which the system approaches steady-state. Expressing the Pollaczek-Khintchine formula with our notation, the mean of the steady-state distribution for server load (often called expected waiting time in the queue, excluding service) is

$$E_z[W(\infty)] = W_F/2, \quad z \geq 0, \rho < 1.$$

4.2. Several Sample Problems

In this section we illustrate the methods described in Section 4.1 by working several sample problems. The first three problems show how changes in the variance of service times affect the transient behavior of mean server load and the steady-state value, assuming the other parameters of the system are held constant. The fourth and fifth problems illustrate using interpolation to determine results for queues when the exact value of ρ or z^* is not tabulated. The determination of waiting

costs using time-dependent mean server load is discussed and applied to one of the sample problems.

The following problems are adapted from an example described by Hillier and Lieberman [1974, p. 436]. A university plans to lease a small batch-processing computer for its students to use. It is expected that the students will submit programs to be run every 3 minutes on the average and that the times between submission of programs have an exponential distribution. The computer under consideration could process an average of 25 typical student programs per hour if it were run continuously. We will examine the transient behavior of this M/G/1 system for an eight-hour period, assuming that the system begins operation with no backlog. Thus far we have specified λ , $E(S)$, and z . For Problem A we assume that the processing times for the students' programs are exponentially distributed, i.e., $k = 1$. Referring to Figure 4.2, we perform the initial computations, using hours as the time unit.

Problem A.

$$\lambda = 20, E(S) = 1/25, k = 1, z = 0$$

$$\rho = 20 \cdot (1/25) = .8$$

$$WF = \frac{20 \cdot 2 \cdot (1/25)^2}{1 - .8} = .32$$

$$EF = .32 / .2 = 1.6$$

$$z^* = 0 / .32 = 0$$

Referring to Figure 4.3 with $\rho = .8$ and $z^* = 0$, we see that Case A is appropriate, i.e., the exact values of ρ and z^* appear in a table in Appendix C. Referring to the appropriate table, we note that the sixteen scaled epochs t^* are between .02 and 4. Eight points should be sufficient to describe the queue's behavior; the selected values of t^* and $E_{z^*}[W^*(t^*)]$ from Appendix C are shown below in the first two columns. Referring to Figure 4.2, we compute t and $E_z[W(t)]$ by multiplying the scaled values times the scaling factors. The desired results are shown below in the last two columns.

Problem A.

t^*	$E_{z^*}[W^*(t^*)]$	t	$E_z[W(t)]$
.1	.164	.16	.052
.2	.230	.32	.074
.4	.306	.64	.098
.6	.351	.96	.112
1.	.405	1.6	.130
2.	.461	3.2	.148
3.	.482	4.8	.154
4.	.491	6.4	.157

The steady-state mean server load in Problem A is $WF/2 = .16$ hour (9.6 minutes), or four average service times. After approximately one hour of operations ($t = .96$), mean server load (.112 hour) in this M/M/1 queue is about 70% of steady-state; after three hours it achieves

90% of the steady-state value. For queues with $\rho < 1$, $E_{z^*}[W^*(t^*)]$ approaches a steady-state value of .5; therefore, the percentage of steady-state achieved at any epoch can be determined simply by doubling the $E_{z^*}[W^*(t^*)]$ value.

For Problem B we consider a system with the same arrival rate and the same mean service time, but we assume that the service times are constant (deterministic). In the context of our original computer-leasing problem, it might be that the student programs are actually uniform or that this specific computer has very little variation in processing times for the programs being run. Referring to Figure 4.2, we perform the computations for this M/D/1 queue.

Problem B.

$$\lambda = 20, E(S) = 1/25, k = \infty, z = 0$$

$$\rho = 20 \cdot (1/25) = .8$$

$$WF = \frac{20 \cdot (1/25)^2}{1 - .8} = .16$$

$$EF = .16 / .2 = .8$$

$$z^* = 0 / .16 = 0$$

Referring to Figure 4.3, we see that the exact values of ρ and z^* are tabulated, and we select eight points from the appropriate table in Appendix C as shown below. The tabulated values are multiplied by EF and WF to obtain t and $E_z[W(t)]$, respectively.

Problem B.

t^*	$E_{z^*}[W^*(t^*)]$	t	$E_z[W(t)]$
.2	.242	.16	.039
.4	.316	.32	.051
.6	.360	.48	.058
1.	.412	.8	.066
1.5	.446	1.2	.071
2.	.465	1.6	.074
3.	.484	2.4	.077
4.	.492	3.2	.079

The steady-state mean server load in Problem B is $WF/2 = .08$ hour (4.8 minutes), or two service times. We observe that after a half-hour of operation ($t = .48$) mean server load (.058 hour) is about 70% of steady-state; after one and a half hours ($t = 1.6$), it has reached 90% of the steady-state value (.074 relative to .08). Compared with the M/M/1 queue of Problem A, this M/D/1 queue has a lower steady-state mean server load and it approaches that equilibrium value much faster.

For Problem C we consider a system with the same parameters as Problems A and B, except that the service times have wide variability. We assume that the mean service time is 2.4 minutes (1/25 hour), as before, but that the standard deviation of the service times is 5.4 minutes. Referring to equation (4.1.1), the appropriate Erlang shape

parameter is $k = .2$. The computations from Figure 4.2 and the selected points from Appendix C are shown below.

Problem C.

$$\lambda = 20, E(S) = 1/25, k = .2, z = 0$$

$$\rho = 20 \cdot (1/25) = .8$$

$$WF = \frac{20 \cdot 6 \cdot (1/25)^2}{1 - .8} = .96$$

$$EF = .96 / .2 = 4.8$$

$$z^* = 0 / .96 = 0$$

t^*	$E_{z^*}[W^*(t^*)]$	t	$E_z[W(t)]$
.1	.156	.48	.150
.2	.223	.96	.214
.4	.299	1.92	.287
.6	.345	2.88	.331
.8	.377	3.84	.362
1.	.400	4.8	.384
1.5	.438	7.2	.420
2.	.459	9.6	.441

The steady-state mean server load in Problem C is $WF/2 = .48$ hour (28.8 minutes), or twelve average service times, which is much higher than the M/M/1 and M/D/1 cases. We observe that about 70% of steady-state is achieved after three hours and that approximately eight hours of operation are needed to reach 90% of the steady-state value.

Figure 4.4 shows mean server load curves for Problems A, B, and C, where these three queues have the same values of λ and $E(S)$. We observe that the queues with higher variance of service times, i.e., with lower values of k , have a higher steady-state mean server load. Furthermore, the approach to equilibrium is slower for the queues with the higher steady-state value. We also note from Figure 4.4 that the ordering of the curves is the opposite of the ordering of the scaled values in the table ($\rho = .8$, $z^* \approx 0$) in Appendix C; this reordering is explained by the different scaling factors, WF and EF , for the three queues.

Returning to our computer-leasing problem, we next consider a computer that can process an average of 30 programs per hour, i.e., $E(S) = 1/30$. We assume that the processing times are exponentially distributed, i.e., $k = 1$, and that the arrival pattern is the same as Problems A, B, and C. For Problem D we will assume that students can leave programs during the night for processing when the facility opens. It is estimated that "about a dozen" programs will arrive during a typical night; thus, the initial server load (assuming 12 jobs at $1/30$ hour each) is .4 hour. Referring to Figure 4.2, we perform the following calculations.

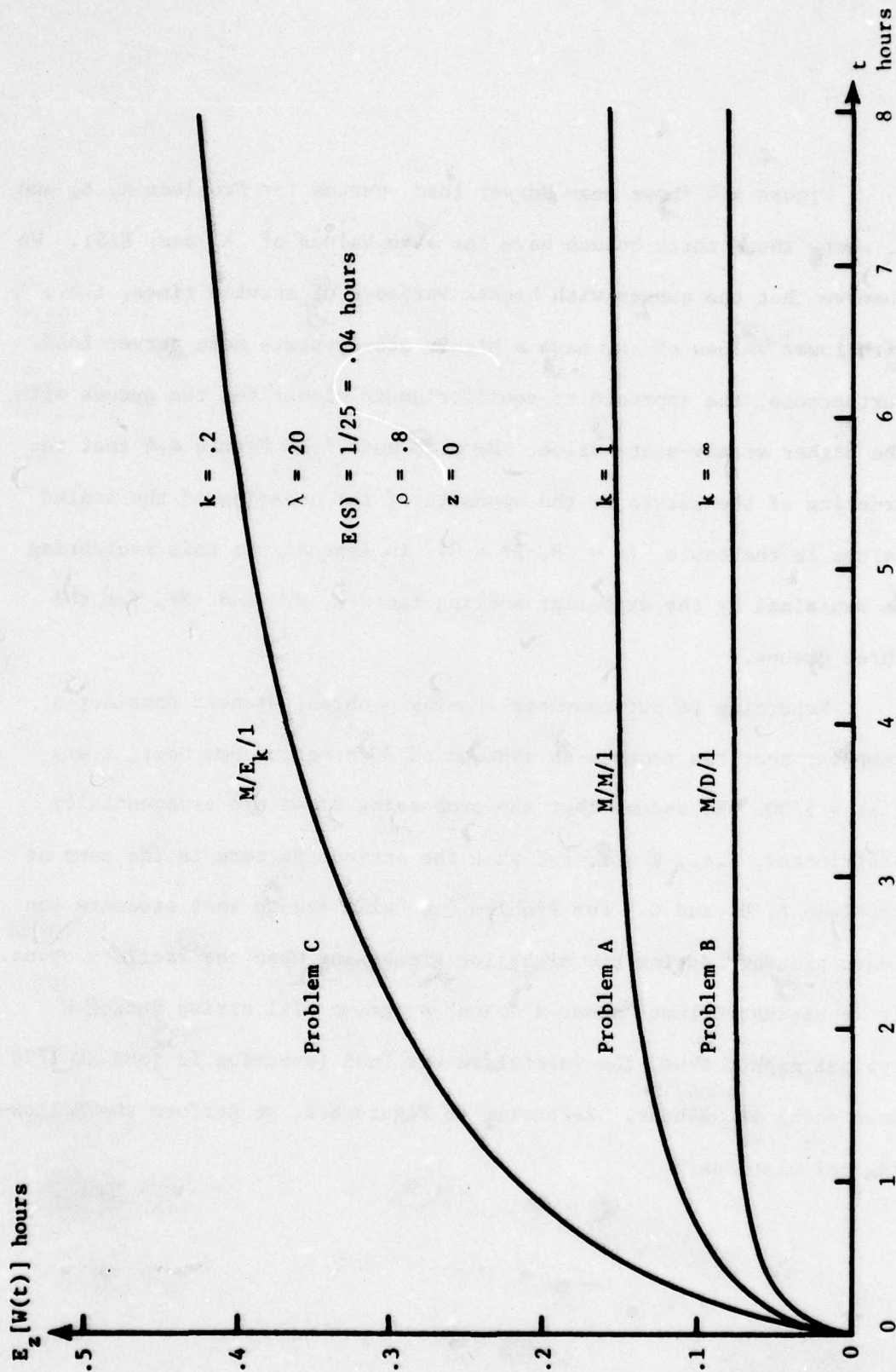


FIGURE 4.4. Graphs for Sample Problems A, B, and C.

Problem D.

Values from Appendix C

$E_{z^*}[W^*(t^*)], z^* = 3$			Interpolated $E_{z^*}[W^*(t^*)]$	$E_z[W(t)]$	
t^*	$\rho = .6$	$\rho = .7$	$\rho = .67$	t	$\rho = .67$
.5	2.5	2.5	2.5	.2	.333
1.	2.0	2.0	2.0	.4	.267
1.5	1.528	1.551	1.543	.6	.206
2.	1.195	1.225	1.215	.8	.162
2.5	.978	1.004	.995	1.	.133
3.	.835	.854	.848	1.2	.113
4.	.673	.681	.678	1.6	.090
5.	.593	.596	.595	2.	.079
6.	.552	.552	.552	2.4	.074
7.	.530	.529	.529	2.8	.071
8.	.517	.517	.517	3.2	.069
10.	.506	.506	.506	4.	.067
12.	.502	.502	.502	4.8	.067

Problem D.

$$\lambda = 20, E(S) = 1/30, k = 1, z = .4$$

$$\rho = 20 \cdot (1/30) = .67$$

$$WF = \frac{20 \cdot 2 \cdot (1/30)^2}{1 - .67} = .133$$

$$EF = .133 / .333 = .4$$

$$z^* = .4 / .133 = 3$$

Referring to Figure 4.3 with $\rho = .67$ and $z^* = 3$, we see that Case B is appropriate. The two relevant tables are $(\rho = .6, z^* = 3)$ and $(\rho = .7, z^* = 3)$, and we list the values of t^* appearing on both tables. Those thirteen values and the corresponding values of $E_{z^*}[W^*(t^*)]$ from the two tables are shown in the first three columns below. For each scaled epoch we interpolate to find the approximate value that is two-thirds (or about seven-tenths) of the difference between the values for $\rho = .6$ and $\rho = .7$. The interpolated value is shown below in the fourth column. Finally, we obtain t and $E_z[W(t)]$ by multiplying t^* and the interpolated values of $E_{z^*}[W^*(t^*)]$ times EF and WF , respectively.

The steady-state mean server load in Problem D is $WF/2 = .067$ hour (4 minutes), or two mean service times. We observe that transient mean server load is within 10% of the steady-state value after approximately two and a half hours of operation. Figure 4.5 shows the curve for Problem D and the curve for the same queue with no initial load. (The calculations for the latter case are not shown here.) The system with $z = 0$ reaches 90% of steady-state after only .8 hour of operation.

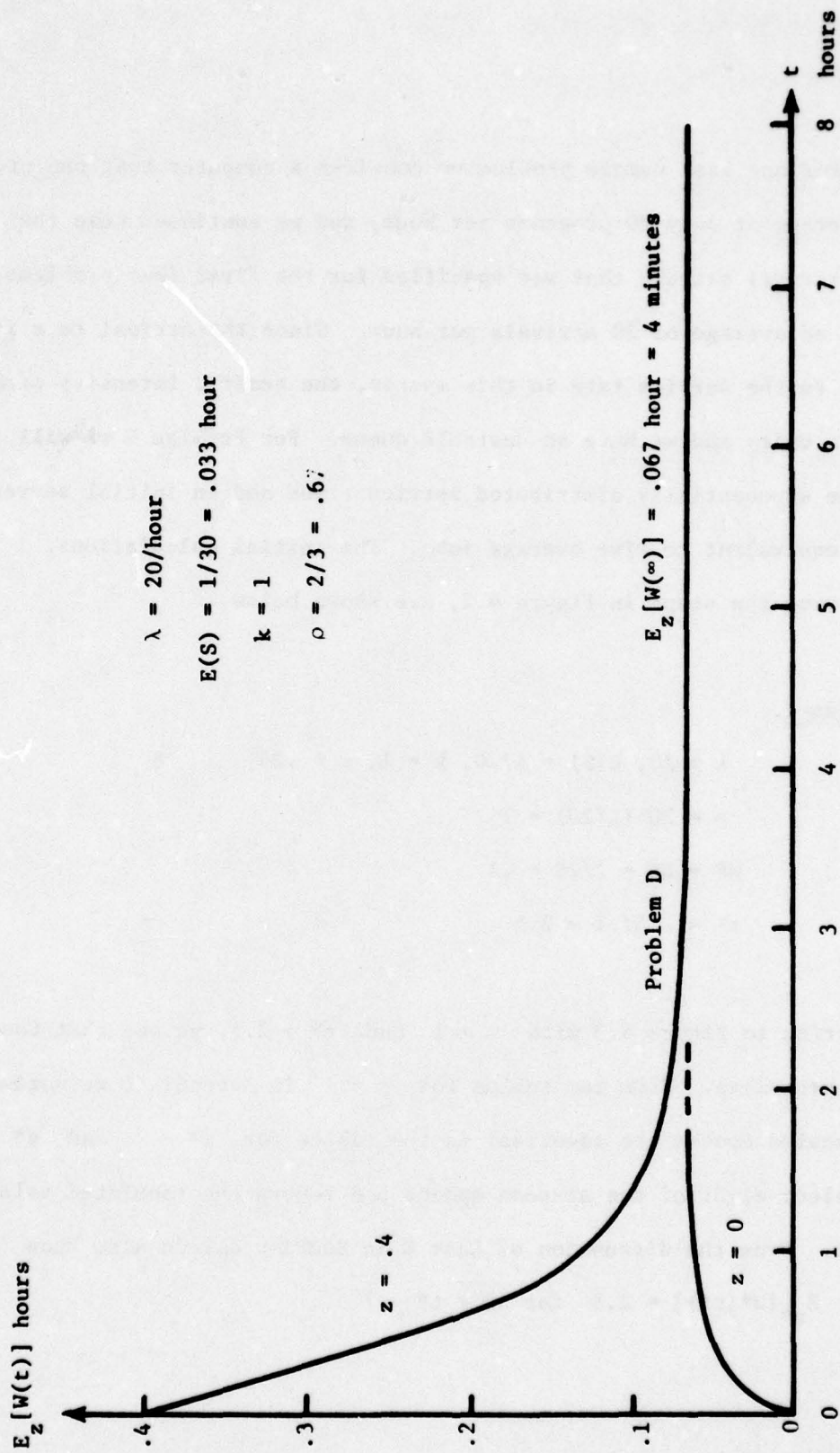


FIGURE 4.5. Graph for Sample Problem D.

For our last sample problem we consider a computer that can process an average of only 20 programs per hour, and we continue to use the same arrival pattern that was specified for the first four problems, i.e., an average of 20 arrivals per hour. Since the arrival rate is equal to the service rate in this system, the traffic intensity parameter equals unity and we have an unstable queue. For Problem E we will assume exponentially distributed service times and an initial server load equivalent to five average jobs. The initial calculations, following the steps in Figure 4.2, are shown below.

Problem E.

$$\lambda = 20, E(S) = 1/20, k = 1, z = .25$$

$$\rho = 20 \cdot (1/20) = 1$$

$$WF = EF = 2/20 = .1$$

$$z^* = .25/.1 = 2.5$$

Referring to Figure 4.3 with $\rho = 1$ and $z^* = 2.5$, we see that Case C is appropriate. From the tables for $\rho = 1$ in Appendix C we note that the scaled epochs are identical in the tables for $z^* = 2$ and $z^* = 3$. We select eight of the sixteen epochs and record the tabulated values below. From the discussion of Case C in Section 4.1 we also know that $E_{z^*}[W^*(t^*)] = 2.5$ for $0 \leq t^* \leq 2.5$.

Problem E.

Values from Appendix C

$E_{z^*}[W^*(t^*)], \rho = 1$			Interpolated $E_{z^*}[W^*(t^*)]$	$E_z[W(t)]$	
t^*	$z^* = 2$	$z^* = 3$	$z^* = 2.5$	t	$z = .25$
0 to 2.5			2.5	0 to .25	.25
3	2.056	3.000	2.528	.3	.253
4	2.149	3.016	2.583	.4	.258
6	2.362	3.108	2.735	.6	.274
8	2.577	3.238	2.908	.8	.291
10	2.784	3.383	3.084	1.	.308
20	3.680	4.115	3.898	2.	.390
40	5.052	5.364	5.208	4.	.521
80	7.068	7.289	7.179	8.	.718

If the M/M/1 system described in Problem E continues operating indefinitely, the mean server load will increase without bound. In many actual systems the arrival rate equals or exceeds the service rate for short periods of time, but such situations may be acceptable if the cost of providing faster service exceeds the cost associated with the high waiting times.

The five sample problems show that the methods described in this chapter can be used to determine time-dependent mean virtual waiting time, an important operating characteristic of M/G/1 queues. In actual applications the analyst may wish to determine the cost

associated with $E_z[W(t)]$ when analyzing or designing a system. In their discussion of decision models for queueing systems, Hillier and Lieberman [1974, Ch. 10] discuss appropriate cost functions for various problems and apply their methods to steady-state distributions of both the number of customers in the system and the waiting times of individual customers. As previously mentioned, our results for expected server load are appropriate for evaluating costs only in systems where the waiting costs are directly proportional to waiting times. We next describe a method for evaluating the waiting cost in situations where the transient behavior of the system is important.

Let c be the cost per unit of waiting time experienced by an individual customer. The cost associated with a customer arriving at epoch t is $c \cdot W(t)$, where there may be some initial server load $W(0) = z$ for process $W(\cdot)$. For any specific epoch t , $W(t)$ is a random variable; in our situation the cost function is linear, and the expected cost for a customer arriving at epoch t is $c \cdot E_z[W(t)]$. In cases where the cost function is non-linear, we would need more information about the distribution of $W(t)$, not just its mean, in order to determine the expected cost of a customer arriving at epoch t . The probability that a customer will arrive during interval $(t, t+\Delta t)$ is $\lambda \Delta t$, and such a customer would incur expected cost $c \cdot E_z[W(t)]$. Thus, the expected waiting cost during a period from $t = 0$ to $t = T$ is

$$(4.2.1) \quad \int_0^T c E_z[W(t)] \lambda dt = c\lambda \int_0^T E_z[W(t)] dt .$$

Note that the integral on the right-hand-side of (4.21) is simply the area under the curve for transient mean server load.

For a queueing system that approaches steady-state quite slowly, the analyst should use the integral expression, not the equilibrium mean waiting time, to evaluate expected cost. To illustrate the difference, consider again the system described in Problem C. Assume the university has determined that for this situation the appropriate cost associated with making a student wait for a program to be processed is \$10 per hour. If we approximate the mean waiting time using the steady-state value of .48 hours, then the expected waiting cost for an individual student arriving with a computer program to be run is \$4.80. The mean arrival rate is 20 per hour, so the expected waiting cost is \$96 for each hour the system is in operation. Thus, during an eight-hour day the total expected waiting cost is \$768. This total cost figure is based on the equilibrium mean waiting time, but from Figure 4.4 we can see that $E_z[W(t)]$ for Problem C is substantially less than the equilibrium value.

For a more accurate determination of total waiting cost during the eight-hour period, we can evaluate the integral on the right-hand-side of equation (4.2.1) graphically by plotting the curve from Figure 4.4 on a fine grid and counting the squares. Using this method the value of the area under the curve is approximately 2.66 hours². (The corresponding value using the equilibrium mean server load is $.48 \cdot 8 = 3.84$ hours².) Multiplying the area under the curve by $c \cdot \lambda$, we find that the correct total expected waiting cost for the eight-hour period is $\$10 \cdot 20 \cdot 2.66 = \532 .

In this particular problem the use of the equilibrium value produces a total cost figure approximately 44% greater than the correct amount. The example illustrates the importance of using time-dependent results when the approach to equilibrium is slow. The graphical technique can also be used to determine expected waiting cost for unstable queueing systems ($\rho \geq 1$), where all behavior is transient and no equilibrium is ever achieved.

CHAPTER 5

CONCLUSIONS

The objective of this research was to provide tables for time-dependent mean server load in M/G/1 queueing systems. Chapter 2 presented the theoretical framework for the server load process, the Laplace transform of $E_z[W(t)]$, and a scaling procedure that allowed us to reduce the number of parameters required to describe a specific system from four $[\lambda, E(S), z, k]$ to three $[\rho, z^*, k]$. In Chapter 3 we considered a technique for inverting the Laplace transform, and we conducted extensive checks and comparisons to ensure the accuracy of our numerical results. The sample problems in Chapter 4 illustrated that a practitioner can easily use our tabulated results in Appendix C by following simple step-by-step procedures. We now discuss some additional results of this research, including a study of the error associated with the Brownian approximation, and we offer some suggestions for future research.

The scaling procedure employed in this research produces curves for $E_{z^*}[W^*(t^*)]$ that are monotonically ordered by each of the three parameters (ρ , z^* , and k), and these consistent orderings facilitated the interpolation procedures described in Chapter 4. Referring to any single table in Appendix C, i.e., fixing the values of ρ and z^* , we observe that the curves for $E_{z^*}[W^*(t^*)]$ increase monotonically as k , the shape parameter of the Erlang service time distribution, increases. (This ordering of the transient curves by the shape parameter has not been proven analytically and is a possible topic for future theoretical

research.) As we would expect, the upper bound (corresponding to $k = \infty$) is the curve for the M/D/1 queue. But our numerical results also indicate that a lower bound can be identified: the gamma input process (corresponding to $k = 0$). We can think of these upper and lower bounds as defining an "envelope", and we might hypothesize that this envelope contains the curves, in scaled form, for all M/G/1 queueing systems, not just for the $M/E_k/1$ queues that are studied here. Figure 5.1 shows the upper and lower bounds for the ($\rho = .5$, $z^* = 0$) case. We observe that the envelope is relatively narrow; specifically, its maximum width is 11% of the steady-state value in this case. For $z^* = 0$ and $\rho = .6, .7, .8$, and $.9$, the maximum width is 9%, 7%, 5%, and 3%, respectively. We could use these bounds to obtain approximate results for any M/G/1 queueing system for which λ , $E(S)$, $E(S^2)$, and z are known, without specifying the exact shape of the service time distribution.

Another monotonic ordering is observed by fixing the values of k and z^* and varying ρ . For example, if we examine the curves for the M/D/1 queue ($k = \infty$) with $z^* = 0$, we observe that for $\rho < 1$ the scaled curves increase monotonically as ρ increases. (In Appendix C this comparison requires examining tables that are each three pages apart.) The M/D/1 curves seem to be approaching the curve for the Wiener process as ρ increases. This observation has not been proven analytically, but the heavy traffic (or Brownian motion) approximation for the equilibrium distribution of M/G/1 server load is derived by examining, the scaled form, the limit of a sequence of queues as ρ

research.) As we would expect, the upper bound (corresponding to $k = \infty$) is the curve for the M/D/1 queue. But our numerical results also indicate that a lower bound can be identified: the gamma input process (corresponding to $k = 0$). We can think of these upper and lower bounds as defining an "envelope", and we might hypothesize that this envelope contains the curves, in scaled form, for all M/G/1 queueing systems, not just for the M/E_k/1 queues that are studied here. Figure 5.1 shows the upper and lower bounds for the ($\rho = .5$, $z^* = 0$) case. We observe that the envelope is relatively narrow; specifically, its maximum width is 11% of the steady-state value in this case. For $z^* = 0$ and $\rho = .6, .7, .8$, and $.9$, the maximum width is 9%, 7%, 5%, and 3%, respectively. We could use these bounds to obtain approximate results for any M/G/1 queueing system for which λ , $E(S)$, $E(S^2)$, and z are known, without specifying the exact shape of the service time distribution.

Another monotonic ordering is observed by fixing the values of k and z^* and varying ρ . For example, if we examine the curves for the M/D/1 queue ($k = \infty$) with $z^* = 0$, we observe that for $\rho < 1$ the scaled curves increase monotonically as ρ increases. (In Appendix C this comparison requires examining tables that are each three pages apart.) The M/D/1 curves seem to be approaching the curve for the Wiener process as ρ increases. This observation has not been proven analytically, but the heavy traffic (or Brownian motion) approximation for the equilibrium distribution of M/G/1 server load is derived by examining, the scaled form, the limit of a sequence of queues as ρ

ρ	t^*						
	.1	.2	.4	.6	1.	2.	4.
2.0	91	60	39	30	21	13	7
1.5	56	38	25	19	14	8	5
1.4	46	32	21	16	12	7	4
1.3	37	25	17	13	10	6	3
1.2	25	18	12	9	7	4	2
1.1	13	9	6	5	4	2	1
.9	14	10	6	4	3	1	0
.8	35	23	14	10	7	3	1
.7	65	42	26	18	12	6	2
.6	108	68	41	30	19	9	3
.5	171	109	67	48	31	15	5

The tabulated error is the difference between the Wiener and the gamma input values, divided by gamma input value, and rounded to the nearest whole per cent.

FIGURE 5.2. Percentage Error of the Wiener Approximation, $z^* = 0$

relationship using a quadratic function. In such cases knowledge of both the first and second moments of the server load distribution is required. Prabhu [1965] derived the double Laplace transform for the time-dependent server load distribution in M/G/1 queues. If we evaluate the second derivative of that expression at zero, we obtain the Laplace transform for the time-dependent second moment of the distribution. The LINV algorithm can then obtain the second moment at specified epochs for use in a quadratic cost function. There may be other queueing applications for this relatively efficient and accurate numerical technique for Laplace transform inversion.

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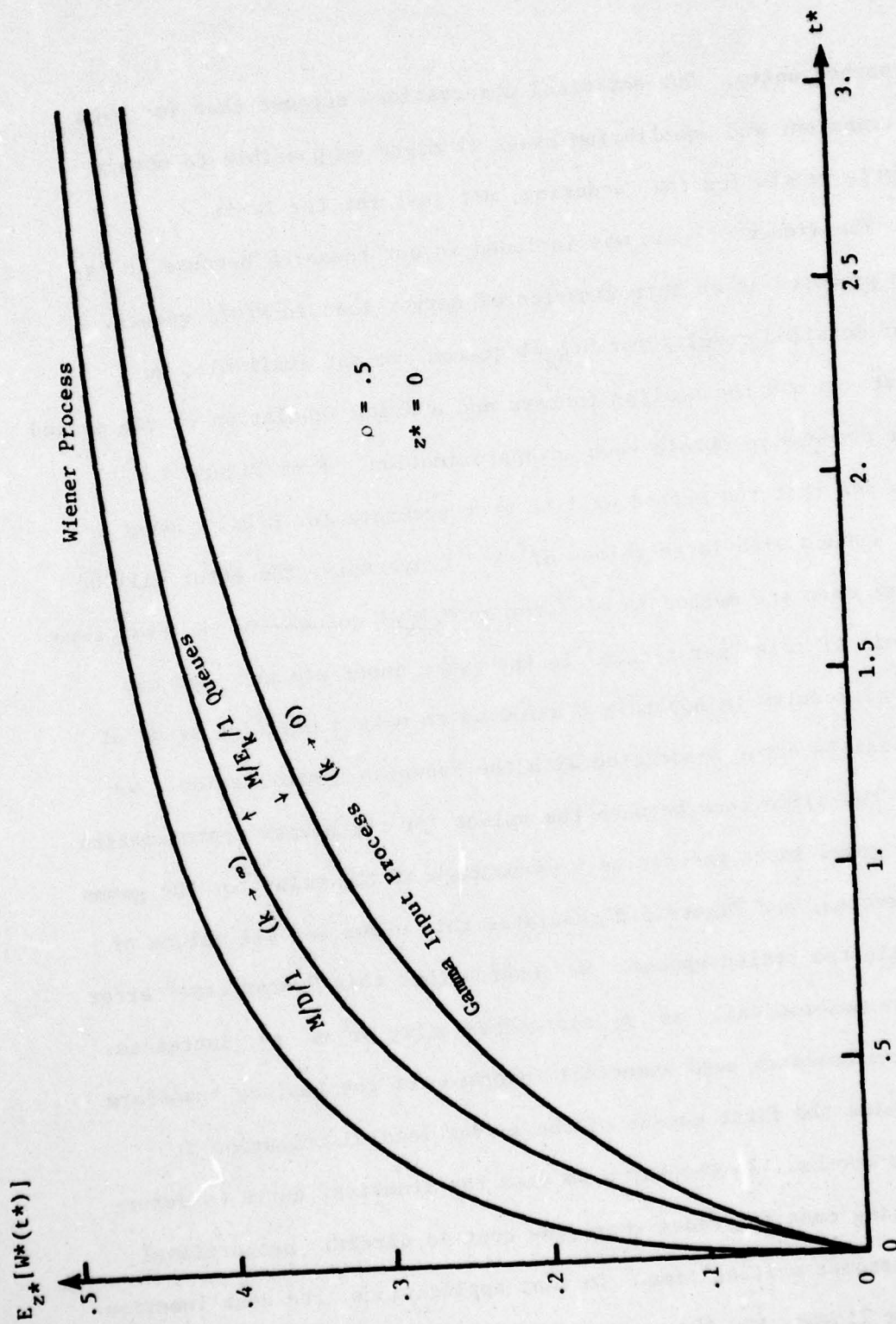


FIGURE 5.1. Several Scaled Processes.

approaches unity. Our empirical observations suggest that for both the transient and equilibrium cases it might be possible to obtain analytic proofs for the ordering, not just for the limit.

The Wiener process was included in our research because it is often proposed as an approximation of server load in $M/G/1$ queues. If our detailed results for $M/E_k/1$ queues are not available, an analyst can use the scaling factors and a brief tabulation of the scaled Wiener process to obtain such an approximation. From Figure 5.1 we can see that the method will be most accurate for $M/D/1$ queues or $M/E_k/1$ queues with large values of k . Conversely, the error will be greatest when the method is used for an $M/E_k/1$ queue with k near zero. The limit of this "worst case" is the gamma input process, and our numerical results in Appendix C allow us to make a detailed study of the possible error associated with the Brownian approximation. We express the difference between the values for the Wiener approximation and the gamma input process as a percentage of the value for the gamma input process, and Figure 5.2 tabulates this error for all values of ρ at selected scaled epochs. We observe that this "worst case" error decreases monotonically as ρ approaches unity or as t^* increases.

This research used numerical inversion of the Laplace transform to determine the first moment of the server load distribution at specified epochs. In Chapter 4 we used the transient curve to determine waiting cost for cases where the cost is directly proportional to the customer waiting time. In many applications, the cost function may not be linear, and the practitioner may approximate the cost

APPENDIX A

SUMMARY OF NOTATION

This list includes brief definitions of the notation used most frequently in this dissertation.

$A(t)$	stationary Poisson process, the M/G/1 arrival process
c	cost per unit of waiting time
$E(\cdot)$	expectation, first moment, of a random variable
$E_z[\cdot]$	expected value, conditional on initial load z
EF	scaling factor for epochs
Fa	an approximation based on Stehfest's algorithm
$F(\cdot)$	cumulative probability distribution for S
$F^*(s)$	Laplace-Stieltjes transform of $F(\cdot)$
$f(\cdot)$	probability density function
$f_n(t; a)$	observational density function with parameters a and n
$G(t)$	gamma process
$g(t)$	correction term function
$H(t)$	$E_z[I(t)] + g(t)$
$I(t)$	cumulative idleness process
k	Erlang shape parameter
N	number of values of the Laplace transform used by Stehfest's algorithm to obtain an approximation
$P(t)$	any function of interest
\bar{P}_n	an approximation of $P(t)$ based on n values of the Laplace transform using Gaver's method

$P_H(z,s)$	Laplace transform of $H(t)$, conditional on z
$P_W(z,s)$	Laplace transform of $W(t)$, conditional on z
$p(s)$	any Laplace transform function
S, S_1	service time random variable
$S(t)$	compound Poisson process, an input process
T	random variable for an observational epoch
$\text{Var}[\cdot]$	variance of a random variable or process
W	random variable for equilibrium server load
WF	scaling factor for waiting time
$W(t)$	server load, or virtual waiting time, process
$X(t)$	net input process, a Levy process
z	deterministic initial server load, $W(0)$
α_i	error components of an asymptotic expansion
α, β	shape and scale parameters for gamma density function
$\Gamma(\cdot)$	gamma function
λ	mean arrival rate, parameter for $A(t)$
μ	$-E[X(t)]/t$
$\xi(t)$	standard Wiener process
ρ	traffic intensity parameter
σ^2	$\text{Var}[X(t)]/t$
$\Phi(\cdot)$	exponent function
$\omega(s)$	solution of functional equation $\Phi[\omega(s)] = s$
-	overscore, or "bar", denotes approximation
*	asterisk superscript, denotes a scaled value

APPENDIX B

LISTINGS OF THE COMPUTER PROGRAMS

1. Figure 3.2, including the documented version of LINV
2. Figure 3.4
3. Figure 3.5
4. Figure 3.6
5. Appendix Tables, $\rho \neq 1$
6. Appendix Tables, $\rho = 1$

COMPUTER PROGRAM FOR FIGURE 3.2

```

-----
      IMPLICIT REAL*8(A-H,O-Z), INTEGER(1-N)
      COMMON/LAFLAC/EPOCH(10),APPROX(10,6)
      DIMENSION EXACT(10,6),STEHP(10,6),V(50)
      EXTERNAL P1,P2,P3,P4,P5,P6
      M=0
      N=34
      DATA STEHP /.56555,.39912,.32655,.28278,
1.25174,.22989,.21322,.19956,.18814,.17796,
1.16566,1.32543,4.47354,
110.60342,20.70645,35.78632,56.82535,84.62735,
1120.78473,165.66749,
1.98775,.91001,.63826,.30968,
1-.02119,-.31527,-.57254,-.76869,-.91049,-.98949,
1-.57762,-1.27084,-1.67544,
1-1.96392,-2.18727,-2.36870,-2.52270,-2.65740,
1-2.77390,-2.88091,
1.36798,.13557,.05043,.01849,
1.00640,.00195,.00036,-.00006,-.00047,-.00020,
1-.66533,-.32531,1.02575,2.39533,
12.78644,1.21092,-3.32956,-11.82953,-25.28353,-44.88511/
      DO 30 I=1,10
      EPOCH(I)=I
      T = I
      EXACT(I,1)=1/DSQRT(1*3.1415926535897932384626433)
      EXACT(I,2)=T*T*T/6
      EXACT(I,3)=DSIN(DSQRT(2*T))
      EXACT(I,4)=-DLOG(T)-0.57721566490153386061
      EXACT(I,5)=DEXP(-T)
      EXACT(I,6)=1-3*T+3*T*T/2-T*T*T/6
      CALL LINV(P1,N,T,APPROX(I,1),V,M)
      CALL LINV(P2,N,T,APPROX(I,2),V,M)
      CALL LINV(P3,N,T,APPROX(I,3),V,M)
      CALL LINV(P4,N,T,APPROX(I,4),V,M)
      CALL LINV(P5,N,T,APPROX(I,5),V,M)
      CALL LINV(P6,N,T,APPROX(I,6),V,M)
30  CONTINUE
      DO 80 MCOPI=1,5
      WRITE(6,4)
      4  FORMAT(1H1,////)
      MARGIN = 4 + MCOPI
      DO 6 IMRGN = 5,MARGIN
      WRITE(6,6)
      6  FORMAT(1H )
      8  CONTINUE
      WRITE(6,40)
      40  FORMAT(1H ,35X,'FIGURE 3.2  INVERSION OF TEST',
1'  FUNCTIONS')
      WRITE(6,42)
      42  FORMAT(1H ,35X,39(' - '))
      WRITE(6,43)
      43  FORMAT(1H0,33X,2(' APPROXIMATE F(T)  USING',21X))

```


COMPUTER PROGRAM FOR FIGURE 3.2

```

      IMPLICIT REAL*8(A-H,O-Z), INTEGER(I-N)
      COMMON/LAFLAC/EPOCH(10),APPROX(10,6)
      LINE&SION EXACT(10,6),STEHP(10,6),V(50)
      EXT&NAL P1,P2,P3,P4,P5,P6
      M=0
      N=34
      DATA STEHP /.56555,.39912,.32655,.28278,
1.25174,.22989,.21322,.19956,.18814,.17796,
1.16566,1.32543,4.47354,
110.60342,20.70845,35.78832,56.82535,84.82735,
1120.78473,165.66749,
1.98775,.91001,.83826,.30988,
1-.02119,-.31527,-.57254,-.76889,-.91049,-.98949,
1-.57762,-1.27084,-1.67544,
1-1.98392,-2.18727,-2.36870,-2.52270,-2.85740,
1-2.77390,-2.88091,
1.36798,.13557,.05043,.01849,
1.00640,.00195,.00036,-.00006,-.00047,-.00020,
1-.66533,-.32531,1.02575,2.39533,
12.78844,1.21092,-3.32956,-11.82953,-25.28353,-44.86511/
      DO 30 I=1,10
      EPOCH(I)=I
      T = I
      EXACT(I,1)=1/DSQ&T(T*3.1415926535897932384626433)
      EXACT(I,2)=T*T*T/6
      EXACT(I,3)=DSIN(DSQ&T(2*T))
      EXACT(I,4)=-DLOG(T)-0.57721566490153386061
      EXACT(I,5)=DEXP(-T)
      EXACT(I,6)=1-3*T+3*T*T/2-T*T*T/6
      CALL LINV(P1,N,T,APPROX(I,1),V,M)
      CALL LINV(P2,N,T,APPROX(I,2),V,M)
      CALL LINV(P3,N,T,APPROX(I,3),V,M)
      CALL LINV(P4,N,T,APPROX(I,4),V,M)
      CALL LINV(P5,N,T,APPROX(I,5),V,M)
      CALL LINV(P6,N,T,APPROX(I,6),V,M)
30    CONTINUE
      DO 80 NCOPY=1,5
      WRITE(6,4)
      4    FORMAT(1H1,////)
      M&GIN = 4 + NCOPY
      DO 8 IM&GN = 5,M&GIN
      WRITE(6,6)
      6    FORMAT(1H )
      8    CONTINUE
      WRITE(6,40)
      40   FORMAT(1H ,35X,'FIGURE 3.2  INVERSION OF TEST',
1'  FUNCTIONS')
      WRITE(6,42)
      42   FORMAT(1H ,35X,39(' '))
      WRITE(6,43)
      43   FORMAT(1H0,33X,2('APPROXIMATE F(T) USING',21X))

```

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F/G 12/1

TRANSIENT EFFECTS IN M/G/1 QUEUES: AN EMPIRICAL INVESTIGATION.(U)

JUN 79 M R MIDDLETON

N00014-76-C-0418

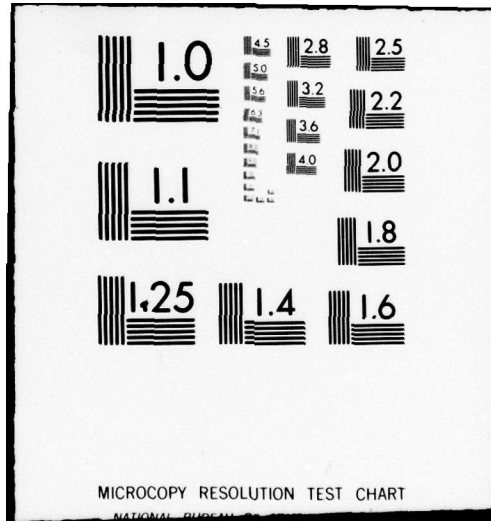
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      WRITE(6,44)
44  FORMAT(1H,35X,2(17HSTEHPFEST'S METHOD,20X))
      WRITE(6,45)
45  FORMAT(1H,32X,2(25(' '),18X))
      WRITE(6,46) N,M
46  FORMAT(1H,11X,'T',6X,2('EXACT F(T)',9X,'M=',12,9X,
1'M=10',7X))
      WRITE(6,48)
48  FORMAT(1H,10X,'----',2(2X,14(' ')),2X,9(' '),
12(2X,14(' ')),2X,9(' '))
      WRITE(6,50)
50  FORMAT(1H0,20X,'F(T) = 1/SQRT(PI*T)',24X,
1'F(T) = -C-LN(T)')
      WRITE(6,52)
52  FORMAT(1H,16X,2(41(' '),2X))
      DO 50 I=1,10
      WRITE(6,54) EPOCH(I), (EXACT(I,J),APPROX(I,J),
1STEHP(I,J), J=1,4,3)
54  FORMAT(1H,10X,F4.1,2(2(2X,F14.10),2X,F9.5))
56  CONTINUE
      WRITE(6,60)
60  FORMAT(1H0,26X,'F(T) = (T*T*T)/6',27X,
1'F(T) = EXP(-T)')
      WRITE(6,52)
      DO 66 I=1,10
      WRITE(6,54) EPOCH(I), (EXACT(I,J),APPROX(I,J),
1STEHP(I,J), J=2,5,3)
66  CONTINUE
      WRITE(6,70)
70  FORMAT(1H0,26X,'F(T) = SIN(SQRT(2*T))',22X,
1'F(T) = 1-3*T+3*T*2/2-T*T*T/6')
      WRITE(6,52)
      DO 76 I=1,10
      WRITE(6,54) EPOCH(I), (EXACT(I,J),APPROX(I,J),
1STEHP(I,J), J=3,6,3)
76  CONTINUE
80  CONTINUE
      STOP
      END
C
C
      SUBROUTINE LINV(P,M,T,FA,V,M)
      IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
      COMMON/INPUT/EPOCH(10),LAMBDA,EXPSVC,RHO,JTOL(8)
      COMMON/OUTPUT/QUAL(10),EXPWT(10,8),ITNTOT(8)
      DIMENSION E(25), F(25), H(25), V(50)
C
C..ARGUMENT P = FUNCTION P(S) = LAPLACE TRANSFORM OF F(T)
C              M = NO. OF VALUES OF S USED TO APPROXIMATE F(T)
C              T = VALUE AT WHICH F(T) IS TO BE APPROXIMATED
C              FA= APPROXIMATE F(T) BASED ON M VALUES OF P(S)
C              V = ARRAY OF COEFFICIENTS 'USED REPEATEDLY'
C              M = FORMAL PARAMETER EXPLAINED BELOW.

```

```

      WRITE(6,44)
44  FORMAT(1H,35X,2(17HSTEHPFEST'S METHOD,20X))
      WRITE(6,45)
45  FORMAT(1H,32X,2(25(' '),18X))
      WRITE(6,46) N,M
46  FORMAT(1H,11X,'T',6X,2('EXACT F(T)',9X,'M=',12,9X,
1'M=10',7X))
      WRITE(6,48)
48  FORMAT(1H,10X,'----',2(2X,14(' ')),2X,9(' '),
12(2X,14(' ')),2X,9(' '))
      WRITE(6,50)
50  FORMAT(1H0,20X,'F(T) = 1/SQRT(PI*T)',24X,
1'F(T) = -C-LN(T)')
      WRITE(6,52)
52  FORMAT(1H,16X,2(41(' '),2X))
      DO 50 I=1,10
      WRITE(6,54) EPOCH(I), (EXACT(I,J),APPROX(I,J),
1STEHP(I,J), J=1,4,3)
54  FORMAT(1H,10X,F4.1,2(2(2X,F14.10),2X,F9.5))
56  CONTINUE
      WRITE(6,60)
60  FORMAT(1H0,26X,'F(T) = (T*T*T)/6',27X,
1'F(T) = EXP(-T)')
      WRITE(6,52)
      DO 66 I=1,10
      WRITE(6,54) EPOCH(I), (EXACT(I,J),APPROX(I,J),
1STEHP(I,J), J=2,5,3)
66  CONTINUE
      WRITE(6,70)
70  FORMAT(1H0,26X,'F(T) = SIN(SQRT(2*T))',22X,
1'F(T) = 1-3*T+3*T*1/2-T*T*T/6')
      WRITE(6,52)
      DO 76 I=1,10
      WRITE(6,54) EPOCH(I), (EXACT(I,J),APPROX(I,J),
1STEHP(I,J), J=3,6,3)
76  CONTINUE
80  CONTINUE
      STOP
      END

```

C
C

```

SUBROUTINE LINV(P,M,T,FA,V,M)
  IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
  COMMON/INPUT/EPOCH(10),LAMBDA,EXPVC,RHO,STOL(8)
  COMMON/OUTPUT/QUAL(10),EXPWT(10,8),ITNTOT(8)
  DIMENSION E(25), F(25), H(25), V(50)

```

C

```

C..ARGUMENT P = FUNCTION P(S) = LAPLACE TRANSFORM OF F(T)
C             M = NO. OF VALUES OF S USED TO APPROXIMATE F(T)
C             T = VALUE AT WHICH F(T) IS TO BE APPROXIMATED
C             FA= APPROXIMATE F(T) BASED ON M VALUES OF P(S)
C             V = ARRAY OF COEFFICIENTS 'USED REPEATEDLY'
C             M = FORMAL PARAMETER EXPLAINED BELOW.

```



```

      IF (M.EQ.N) GO TO 50
C
C..ON FIRST CALL OF LINV (M NOT EQUAL TO N) ARRAY V(I) MUST
C BE EVALUATED.
C
C..FIRST COMPUTE E(I) = FACTORIAL OF 2I OVER THOSE OF I AND
C I-1, AND F(I) = FACTORIAL OF I-1.
C
      NH = N/2
      E(1) = 2
      E(2) = 12
      F(1) = 1
      F(2) = 1
C
      DO 10 I = 2, NH
      E(I+1) = ( 2.DO*(2*I+1)*E(I) ) / I
      F(I+1) = F(I)*I
10  CONTINUE
C
C..NEXT COMPUTE H(J) = ALL TERMS INDEPENDENT OF INDEX I IN
C THE V(I) SUMMATION.
C
      PWR = NH
      H(1) = 2/F(NH)
C
      DO 20 J = 2, NH
      H(J) = (J**PWR)*E(J)/F(NH-J+1)
20  CONTINUE
C
C..FINALLY EVALUATE ARRAY V(I).
C
      ISIGN = 2*(NH-2*(NH/2))-1
C
      DO 40 I = 1, N
      V(I) = 0
      JFIRST = (I+1)/2
      JLAST = MIN0(I,NH)
C
      DO 30 J = JFIRST, JLAST
      V(I) = V(I) + H(J)/(F(I-J+1)*F(2*J-I+1))
30  CONTINUE
C
      V(I) = ISIGN*V(I)
      ISIGN = -ISIGN
40  CONTINUE
C
      M = N
C
C..SUBSEQUENT CALLS (M EQUAL TO N) WILL USE ARRAY V(I) FROM
C COMMON STORAGE AND WILL BRANCH TO THIS SECTION OF LINV.
C
50  PA = 0
      A = .69314718055994530941723212145880 / 1

```



```

        DO 60 I= 1, N
        FA = FA + V(I)*P(I*A)
60      CONTINUE
C
        FA = A*FA
C
        RETURN
        END
C
C
        DOUBLE PRECISION FUNCTION P1(S)
        IMPLICIT REAL*8(A-H,O-Z), INTEGER(I-N)
        F1=1/DSQRT(S)
        RETURN
        END
C
C
        DOUBLE PRECISION FUNCTION P2(S)
        IMPLICIT REAL*8(A-H,O-Z), INTEGER(I-N)
        F2=1/(S*S*S*S)
        RETURN
        END
C
C
        DOUBLE PRECISION FUNCTION P3(S)
        IMPLICIT REAL*8(A-H,O-Z), INTEGER(I-N)
        F3=DSQRT(3.1415926535897932384626433/(2*S*S*S))
        F3=DEXP(-1/(2*S))
        RETURN
        END
C
C
        DOUBLE PRECISION FUNCTION P4(S)
        IMPLICIT REAL*8(A-H,O-Z), INTEGER(I-N)
        F4=DLOG(S)/S
        RETURN
        END
C
C
        DOUBLE PRECISION FUNCTION P5(S)
        IMPLICIT REAL*8(A-H,O-Z), INTEGER(I-N)
        F5=1/(S+1)
        RETURN
        END
C
C
        DOUBLE PRECISION FUNCTION P6(S)
        IMPLICIT REAL*8(A-H,O-Z), INTEGER(I-N)
        F6=(S-1)*(S-1)*(S-1)/(S*S*S*S)
        RETURN
        END

```

COMPUTER PROGRAM FOR FIGURE 3.4

```

-----
      IMPLICIT REAL*8(A-H,O-Z), INTEGER(I-N)
      COMMON PA1(10),PA2(9)
      DIMENSION F1(10),S1(10),D1(10),V1(10),
1F2(9),B2(9),V2(9),V(50),T2(9)
      EXTERNAL F1,F2
      N=0
      M=34
      DO 20 I=1,10
      T=I
      F1(I)=1/DSQRT(T*3.1415926535897932384626433)
      CALL LINV(P1,M,T,FA1(I),V,M)
20  CONTINUE
C
      DATA S1/.56555,.39912,.32655,.28276,
1.25174,.22989,.21322,.19956,.18614,.17796/
      DATA D1/.73172,.40035,.26343,.26286,
1.29365,.22901,.18062,.20112,.21609,.17650/
      DATA V1/.56419,.39694,.32573,.28209,
1.25231,.23033,.21324,.19947,.18606,.17841/
      DATA T2/4.140186,2.501126,1.643438,1.085064,
1.693147,.412298,.214821,.085541,.016048/
C
      DO 30 I=1,9
      T=T2(I)
      F2(I)=DEXP(-T/2)
      CALL LINV(P2,M,T,FA2(I),V,M)
30  CONTINUE
C
      DATA B2/.120527,.288195,.439064,.581306,
1.707316,.813401,.898482,.957847,.992205/
      DATA V2/.126174,.286329,.439675,.581265,
1.707107,.813712,.898156,.958135,.992015/
C
      DO 180 NCOPY=1,5
      WRITE(6,4)
4  FORMAT(1H1,////)
      MARGIN = 4 + NCOPY
      DO 8 IMRGH = 5,MARGIN
      WRITE(6,6)
6  FORMAT(1H )
8  CONTINUE
C
      WRITE(6,40)
40  FORMAT(1H ,33X,'FIGURE 3.4  COMPARISONS',
1' WITH OTHER TECHNIQUES')
      WRITE(6,48)
48  FORMAT(1H ,33X,45(' - '))
      WRITE(6,52)
52  FORMAT(1H0,20X,'F(T) = 1/SQRT(T*PI)')

```



```

      WRITE(6,54)
54  FORMAT(1H ,20X,19('-'),3X,'APPROXIMATE F(T) ',
1' USING NUMERICAL INVERSION')
      WRITE(6,60)
60  FORMAT(1H ,42X,42('-'))
      WRITE(6,64)
64  FORMAT(1H ,44X,17HSTEINFEST'S METHOD)
      WRITE(6,66)
66  FORMAT(1H ,42X,21('-'))
      WRITE(6,70) M
70  FORMAT(1H ,22X,'T      EXACT F(T)      M=',12,6X,
1'M=10*      DUBNE*      VEILLON*')
      WRITE(6,72)
72  FORMAT(1H ,21X,'---',2(4X,10('-')),3(4X,7('-'))).
      DO 78 I=1,10
      WRITE(6,76) I,F1(I),FA1(I),S1(I),D1(I),V1(I)
76  FORMAT(1H ,21X,12,5X,2(F10.6,4X),3(F7.5,4X))
78  CONTINUE

C
      WRITE(6,80)
80  FORMAT(1H-,20X,'F(T) = EXP(-T/2) ')
      WRITE(6,82)
82  FORMAT(1H ,20X,16('-'),10X,'APPROXIMATE F(T) ',
1' USING NUMERICAL INVERSION')
      WRITE(6,85)
85  FORMAT(1H ,46X,42('-'))
      WRITE(6,86) M
86  FORMAT(1H ,24X,'T',9X,'EXACT F(T)',6X,
1'STEINFEST M=',12,4X,'BELLMAN*      VEILLON*')
      WRITE(6,88)
88  FORMAT(1H ,20X,8('-'),2(3X,15('-')),2(3X,8('-'))).
      DO 94 I=1,9
      WRITE(6,92) I2(I),F2(I),FA2(I),B2(I),V2(I)
92  FORMAT(1H ,20X,F8.6,2(3X,F15.13),2(3X,F8.6))
94  CONTINUE

C
      WRITE(6,102)
102 FORMAT(1H-,25X,'*NOTE: DATA FOR METHODS MARKED',
1' BY AN ASTERISK WERE TAKEN')
      WRITE(6,104)
104 FORMAT(1H ,32X,'FROM A.C.M ALGORITHM 486, ',
123H'NUMERICAL INVERSION OF)
      WRITE(6,106)
106 FORMAT(1H ,32X,23H'LAFLACE TRANSFORM', BY ,
1'FRANCOISE VEILLON, COMMUNI-')
      WRITE(6,108)
108 FORMAT(1H ,32X,'CATIONS OF THE A.C.M.,',
1' VOLUME 17, NUMBER 10,')
      WRITE(6,110)
110 FORMAT(1H ,32X,'OCTOBER 1974.')
```

180 CONTINUE

STOP

END


```

SUBROUTINE LINV(P,N,T,FA,V,d)
  IMPLICIT REAL*8(A-H,O-Z), INTEGER(I-N)
  COMMON FA1(10),FA2(9)
  DIMENSION E(25), F(25), H(25), V(50)
  IF (A.EQ.0) GO TO 50
  MH = M/2
  E(1) = 2
  E(2) = 12
  F(1) = 1
  F(2) = 1
  DO 10 I = 2, MH
    E(I+1) = ( 2.DO*(2*I+1)*E(I) ) / I
    F(I+1) = F(I)*I
10  CONTINUE
    PMH = MH
    H(1) = 2/F(MH)
    DO 20 J = 2, MH
      H(J) = ((J**PMH)*E(J))/F(MH-J+1)
20  CONTINUE
    ISIGN = 2*(MH-2*(MH/2))-1
    DO 40 I = 1, N
      V(I) = 0
      JFIRST = (I+1)/2
      JLAST = MIN0(I,MH)
      DO 30 J = JFIRST, JLAST
        V(I) = V(I) + H(J)/(F(I-J+1)*F(2*J-I+1))
30  CONTINUE
      V(I) = ISIGN*V(I)
      ISIGN = -ISIGN
40  CONTINUE
    M = N
50  FA = 0
    A = .69314718055994530941723212145880 / T
    DO 60 I = 1, N
      FA = FA + V(I)*P(I*A)
60  CONTINUE
    FA = A*FA
    RETURN
  END

```

C
C

```

DOUBLE PRECISION FUNCTION P1(S)
  IMPLICIT REAL*8(A-H,O-Z), INTEGER(I-N)
  P1=1/DSQRT(S)
  RETURN
  END

```

C
C

```

DOUBLE PRECISION FUNCTION P2(S)
  IMPLICIT REAL*8(A-H,O-Z), INTEGER(I-N)
  P2=2/(1+2*S)
  RETURN
  END

```

COMPUTER PROGRAM FOR FIGURE 3.5

```

      IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
      COMMON/INPUT/EPOCH(10),LAMBDA,EXPSVC,RHO,JTOL(8)
      COMMON/OUTPUT/QUAL(10),EXPWT(10,8),ITNTOT(8)
      COMMON/LAPLAC/V(50),M,N
      COMMON/MM1/OSTART,SS
      COMMON/NEWRAP/TOLRNC,ITNSUM
      EXTERNAL EQUAD,PMH1
      EXTERNAL FMM1,DFMM1,DDFMM1
      N = 0
      N = 34
C      Z = 0.00
C
      DO 20 I = 1,10
      CALL LINV(EQUAD,M,EPOCH(I),EZIT,V,M)
      QUAD(I) = (RHO-1.)*EPOCH(I) + EZIT
20    CONTINUE
C
      DO 50 J=1,8
      JTOL(J) = 2*J + 8
      TOLRNC = 1./(10.**JTOL(J))
      OSTART = 1
      ITNSUM = 0
C
      DO 40 I=1,10
      CALL LINV(PMH1,M,EPOCH(I),EZIT,V,M)
      EXPWT(I,J) = (RHO-1.)*EPOCH(I) + EZIT
40    CONTINUE
C
      ITNTOT(J) = ITNSUM
C
C      50 CONTINUE
C
      CALL TOLOUT
C
      STOP
      END
C
C
      BLOCK DATA
      IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
      COMMON/INPUT/EPOCH(10),LAMBDA,EXPSVC,RHO,JTOL(8)
      DATA EPOCH /20.,40.,60.,80.,100.,
1200.,300.,400.,800.,1200./
      DATA LAMBDA,EXPSVC,RHO /1.00,0.95,0.95/
      END
C
C
      SUBROUTINE TOLOUT
      IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
      COMMON/INPUT/EPOCH(10),LAMBDA,EXPSVC,RHO,JTOL(8)
      COMMON/OUTPUT/QUAL(10),EXPWT(10,8),ITNTOT(8)

```



```

COMMON/LAPLAC/V(50),B,M
DIMENSION BESSEL(5),JTOL(8)

C
BESSEL(1) = 3.9161
BESSEL(2) = 5.4651
BESSEL(3) = 6.5582
BESSEL(4) = 7.4191
BESSEL(5) = 8.1335

C
DO 8C NCOPY=1,5
WRITE(6,4)
4 FORMAT(1H1,////)
MARGIN = 4 + NCOPY
DO 8 IMGM = 5,MARGIN
WRITE(6,6)
6 FORMAT(1H )
8 CONTINUE

C
WRITE(6,12)
12 FORMAT(1H ,29X,'FIGURE 3.5 EFFECT OF NEWTON-RAPHSON',
1' SEARCH TOLERANCE')

C
WRITE(6,14)
14 FORMAT(1H ,29X,53('-'))

C
WRITE(6,16)
16 FORMAT(1H0,10X,37HTHIS TABLE COMPARES COLEMAN'S BESSEL ,
1'FUNCTION RESULTS AT FIVE EPOCHS WITH LAPLACE')

C
WRITE(6,18)
18 FORMAT(1H ,10X,'INVERSION RESULTS FOR BOTH THE EXACT ',
1'QUADRATIC SOLUTION OF THE FUNCTIONAL AND THE ')

C
WRITE(6,20)
20 FORMAT(1H ,10X,'APPROXIMATE NEWTON-RAPHSON SEARCH ',
1'SOLUTION OF THE FUNCTIONAL WITH VARIOUS TOLERANCES.')
WRITE(6,30)
30 FORMAT(1H0,10X,'M/M/1 MEAN SERVER LOAD,',
1' LAMBDA = 1.0, E(S) = 0.95, Z = 0.')

C
DO 8C JFIRST=1,5,4
JLAST=JFIRST+3

C
WRITE(6,36)
36 FORMAT(1H0,36X,'INVERSION OF LAPLACE TRANSFORM,',
122H STEHFEST'S METHOD, N=,12)

C
WRITE(6,40)
40 FORMAT(1H ,18X,'SUMS OF ',73('-'))

C
WRITE(6,42)
42 FORMAT(1H ,19X,'BESSEL QUADRATIC TOLERANCE FOR ',
1'NEWTON-RAPHSON SEARCH SOLUTION OF FUNCTIONAL')

```



```

C      WRITE(6,44)
44  FORMAT(1H ,10X,'EPOCH FUNCTIONS SOLUTION OF ',58(' '))
C
      DO 46 J=JFIRST,JLAST
46  CONTINUE
C
      WRITE(6,50) (JTOL(J),J=JFIRST,JLAST)
50  FORMAT(1H ,10X,' T      (COLEMAN) FUNCTIONAL      ',
14('10**-',12,8X))
C
      WRITE(6,56)
56  FORMAT(1H ,10X,'-----      -----',5(2X,13(' ')))
C
      DO 62 I=1,5
C
      WRITE(6,60) EPOCH(I),BESSEL(I),QUAD(I),
1(EXPWT(I,J),J=JFIRST,JLAST)
60  FORMAT(1H ,10X,F5.0,3X,F7.4,5(1X,F14.11))
62  CONTINUE
C
      DO 66 I=6,10
C
      WRITE(6,64) EPOCH(I),QUAD(I),(EXPWT(I,J),J=JFIRST,JLAST)
64  FORMAT(1H ,10X,F5.0,10X,5(1X,F14.11))
C
66  CONTINUE
C
      WRITE(6,70)
70  FORMAT(1H ,10X,'TOTAL NUMBER OF NEWTON-RAPHSON')
C
      WRITE(6,72) (INTOT(J),J=JFIRST,JLAST)
72  FORMAT(1H ,10X,'ITERATIONS TO EVALUATE 10 EPOCHS: ',
14(14,11X))
C
80  CONTINUE
      RETURN
      ENL
C
C
SUBROUTINE LINV(P,N,T,FA,V,a)
IMPLICIT REAL*8 (A-H,O-Z), INTEGER(I-N)
COMMON/LAPLAC/EPOCH(10),APPROX(10,6)
DIMENSION E(25), F(25), H(25), V(50)
IF (N.EQ.0) GO TO 50
NH = N/2
E(1) = 2
E(2) = 12
F(1) = 1
F(2) = 1
DO 10 I = 2, NH
E(I+1) = ( 2.00*(2*I+1)*E(I) ) / I
F(I+1) = F(I)*I

```

```

10  CONTINUE
    FWH = NH
    H(1) = 2/F(NH)
    DO 20 J = 2, NH
        H(J) = ((J*FWH)*E(J))/F(NH-J+1)
20  CONTINUE
    ISIGN = 2*(NH-2*(NH/2))-1
    DO 40 I = 1, N
        V(I) = 0
        JFIRST = (I+1)/2
        JLAST = MIN0(I,NH)
        DO 30 J = JFIRST, JLAST
            V(I) = V(I) + H(J)/(F(I-J+1)*F(2*J-I+1))
30  CONTINUE
        V(I) = ISIGN*V(I)
        ISIGN = -ISIGN
40  CONTINUE
    N = N
50  FA = C
    A = .693147180559945309417232121458DC / T
    DO 60 I = 1, N
        FA = FA + V(I)*P(I*A)
60  CONTINUE
    FA = A*FA
    RETURN
    END

```

C
C

```

DOUBLE PRECISION FUNCTION PQUA(S)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON/INPUT/EPOCH(10),LAMBDA,EXPSVC,RHC,JTOL(8)
B = 1 - EXPSVC*(LAMBDA+S)
OMEGA = (-B+LSQRT(B*B+4*S*EXPSVC))/(2*EXPSVC)
PQUA=DEXP(-OMEGA*Z)/(S*OMEGA)
RETURN
END

```

C

```

DOUBLE PRECISION FUNCTION PMH1(S)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON/INPUT/EPOCH(10),LAMBDA,EXPSVC,RHO,JTOL(8)
COMMON/MH1/OSTART,SS
EXTERNAL FCHM1,DFHM1
SS = S
CALL ZERO(FCHM1,DFHM1,OSTART,OMEGA)
OSTART = OMEGA
PMH1=DEXP(-OMEGA*Z)/(S*OMEGA)
RETURN
END

```

C
C

```

DOUBLE PRECISION FUNCTION PMH1(S)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON/INPUT/EPOCH(10),LAMBDA,EXPSVC,RHC,JTOL(8)

```

```

FMH1 = S - LAMBDA*(1 - (1/(1+S*EXPSVC)))
RETURN
END

```

C
C

```

DOUBLE PRECISION FUNCTION DFHM1(S)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON/INPUT/EPOCH(10),LAMBDA,EXPSVC,RHO,JTOL(8)
DFHM1 = 1 - LAMBDA*EXPSVC*(1/(1+S*EXPSVC))**2
RETURN
END

```

C
C

```

DOUBLE PRECISION FUNCTION DDFHM1(S)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON/INPUT/EPOCH(10),LAMBDA,EXPSVC,RHO,JTOL(8)
DDFH1 = 2*LAMBDA*EXPSVC**2*(1/(1+S*EXPSVC))**3
RETURN
END

```

C

```

DOUBLE PRECISION FUNCTION FORM1(OMEGA)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON/MH1/OSTART,SS
FORM1 = FMH1(OMEGA) - SS
RETURN
END

```

C
C

```

SUBROUTINE ZERO(F,DF,XSTART,XZERO)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON/NEWHAF/TOLRNC,ITNSUM
ITERAT = 0
X = XSTART
10 CONTINUE
XNEW = X - F(X)/DF(X)
RELACC = (XNEW-X)/XNEW
X = XNEW
ITERAT = ITERAT + 1
IF (ITERAT .GE. 50) GO TO 50
IF (DABS(RELACC) .LE. TOLRNC) GO TO 20
GO TO 10
20 CONTINUE
ITNSUM = ITNSUM + ITERAT
XZERO = X
RETURN

```

C

```

50 CONTINUE
WRITE(6,55) RELACC, TOLRNC
55 FORMAT(1H, 'ITERAT=50, RELACC=', D24.16, ', TOLRNC=', D24.16)
ITNSUM = ITNSUM + ITERAT
XZERO = X
RETURN
END

```


COMPUTER PROGRAM FOR FIGURE 3.6

```

      IMPLICIT REAL*8 (A-E,O-Z), INTEGER(I-N)
      COMMON /RHOZ/ RHO,Z
      DIMENSION SEZIT1(11,8),SEZIT2(11,8),V(50),ZSIN(8),RHOIN(11)
      EXTERNAL PQUAD1,PQUAD2
      DATA ZSIN /.2,.4,.6,.8,1.,2.,3.,4./
      DATA RHOIN /.5,.6,.7,.8,.9,1.1,1.2,1.3,1.4,1.5,2./

C
      M=0
      N=34

C
      DO 12 I=1,11
      RHO = RHOIN(I)
      WTFCTR = LABS(1.-RHO)/RHO
      DO 12 JZ=1,8
      Z = ZSIN(JZ)/WTFCTR
      CALL LINV(PQUAD1,M,Z,EZIZ,V,M)
      SEZIT1(I,JZ) = EZIZ*WTFCTR
      CALL LINV(PQUAD2,M,Z,EZIZ,V,M)
      SEZIT2(I,JZ) = EZIZ*WTFCTR
12  CONTINUE

C
      DO 88 NCOPY=1,5
      WRITE(6,20)
20  FORMAT(1H1,////)
      MARGIN = 4 + NCOPY
      DO 28 I=1,MARGIN
      WRITE(6,24)
24  FORMAT(1H )
28  CONTINUE

C
      WRITE(6,32)
32  FORMAT(1H ,26X,'FIGURE 3.6 CORRECTIONS',
1' FOR DISCONTINUOUS FIRST DERIVATIVE')
      WRITE(6,36)
36  FORMAT(1H ,26X,56(' '))
      WRITE(6,40) M
40  FORMAT(1H-,33X,'APPROXIMATIONS USING',
122H STEPHENSON'S METHOD, M=,12)
      WRITE(6,44)
44  FORMAT(1H-,20X,'E/H/1 SCALED MEAN CUMULATIVE',
141H IDLENESS AT T'=ABS(MU)*Z' (I.E., AT T=Z))
      WRITE(6,48)
48  FORMAT(1H0,47X,'WITHOUT CORRECTION TERM')
      WRITE(6,52)
52  FORMAT(1H0,45X,'SCALED INITIAL SERVER LOAD')
      WRITE(6,56)
56  FORMAT(1H ,19X,76(' '))
      WRITE(6,60) (ZSIN(JZ), JZ=1,8)
60  FORMAT(1H ,15X,'RHO',5X,8(F3.1,7X))
      WRITE(6,64)
64  FORMAT(1H ,15X,'---',8(2X,8(' ')))

```

```

DO 72 IRHO=1,11
WRITE(6,68) RHOIN(IRHO), (SEZIT1(IRHO,JZ), JZ=1,8)
68 FORMAT(1H,15X,F3.1,8(2X,F8.0))
72 CONTINUE
C
WRITE(6,76)
76 FORMAT(1H)
WRITE(6,44)
WRITE(6,80)
80 FORMAT(1H0,48X,'WITH CORRECTION TERM')
WRITE(6,52)
WRITE(6,56)
WRITE(6,60) (ZSIN(JZ), JZ=1,8)
WRITE(6,64)
DO 84 IRHO=1,11
WRITE(6,68) RHOIN(IRHO), (SEZIT2(IRHO,JZ), JZ=1,8)
84 CONTINUE
88 CONTINUE
STOP
END

```

C
C

```

SUBROUTINE LINV(P,M,T,FA,V,M)
IMPLICIT REAL*8(A-H,O-Z), INTEGER(I-N)
COMMON/LAPLAC/EPOCH(10),APPROX(10,6)
DIMENSION E(25), F(25), H(25), V(50)
IF (M.EQ.N) GO TO 50
NH = M/2
E(1) = 2
E(2) = 12
F(1) = 1
F(2) = 1
DO 10 I = 2, NH
E(I+1) = ( 2.DO*(2*I+1)*E(I) ) / I
F(I+1) = F(I)*I
10 CONTINUE
PWH = NH
H(1) = 2/F(NH)
DO 20 J = 2, NH
H(J) = ((J**PWH)*E(J))/F(NH-J+1)
20 CONTINUE
ISIGN = 2*(NH-2*(NH/2))-1
DO 40 I = 1, N
V(I) = 0
JFINST = (I+1)/2
JLAST = MIN0(I,NH)
DO 30 J = JFINST, JLAST
V(I) = V(I) + H(J)/(F(I-J+1)*F(2*J-I+1))
30 CONTINUE
V(I) = ISIGN*V(I)
ISIGN = -ISIGN
40 CONTINUE
N = M

```

```

50  FA = 0
    A = .29314718055994530941723212145800 / 1
    DO 60 I= 1, N
      FA = FA + V(I)*P(I*FA)
60  CONTINUE
    FA = A*FA
    RETURN
    END

```

C
C

```

DOUBLE PRECISION FUNCTION PQUAD1(S)
IMPLICIT REAL*8(A-H,O-Z), INTEGER(I-N)
COMMON /RHOZ/ RHO,Z
B = 1 - RHO - S/2.
OMEGA = -B+DSQRT(B*B+2*S)
PQUAL1 = DEXP(-OMEGA*Z)/(S*OMEGA)
RETURN
END

```

C
C

```

DOUBLE PRECISION FUNCTION PQUAD2(S)
IMPLICIT REAL*8(A-H,O-Z), INTEGER(I-N)
COMMON /RHOZ/ RHO,Z
B = 1 - RHO - S/2.
OMEGA = -B+DSQRT(B*B+2*S)
PQUAD2=DEXP(-OMEGA*Z)/(S*OMEGA)-DEXP(-Z*(2.*RHO+S))/(S*S)
RETURN
END

```


COMPUTER PROGRAM FOR APPENDIX TABLES, RHO NOT EQUAL TO ONE

```

C
      IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
      COMMON /INPUT/ TVAL(16),KVAL(7),SCLWT(10,16,3),RHOIN,
1ZSIN(3)
      COMMON /LAPLAC/ V(50),A,M
      COMMON /NEWRAP/ TOLRMC,MAXITN
C
      M=0
      N=34
      MAXITN=20
      TOLRMC=1.D-20
C
      DO 60 IFAGE=1,3
      READ(5,20) RHOIN, (ZSIN(ITABLE), ITABLE=1,3)
      READ(5,20) (TVAL(J), J=1,6)
      READ(5,20) (TVAL(J), J=7,12)
      READ(5,20) (TVAL(J), J=13,16)
20  FORMAT(6F10.0)
C
      DO 50 ITABLE=1,3
      CALL RZWNA(ITABLE)
      CALL RZMD1(ITABLE)
      CALL RZNEK(ITABLE)
      CALL RZGAN(ITABLE)
50  CONTINUE
      CALL RZOUT
60  CONTINUE
      STOP
      END
C
C
      BLOCK DATA
      IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
      COMMON /INPUT/ TVAL(16),KVAL(7),SCLWT(10,16,3),RHOIN,
1ZSIN(3)
      DATA KVAL/4.,3.,2.,1.5,1.,.5,.2/
      END
C
C
      SUBROUTINE LINV(P,M,T,FA,V,M)
      IMPLICIT REAL*8(A-H,O-Z), INTEGER(I-N)
      COMMON/LAPLAC/EPOCH(10),APPROX(10,6)
      DIMENSION E(25), F(25), H(25), V(50)
      IF (M.EQ.N) GO TO 50
      NH = M/2
      E(1) = 2
      E(2) = 12
      F(1) = 1
      F(2) = 1
      DO 10 I = 2, NH
      E(I+1) = ( 2.00*(2*I+1)*E(I) ) / I
      F(I+1) = F(I)*I

```

```

10  CONTINUE
    PWB = MH
    H(1) = 2/F(MH)
    DO 20 J = 2, MH
        H(J) = ((J**PWB)*E(J))/F(MH-J+1)
20  CONTINUE
    ISIGN = 2*(MH-2*(MH/2))-1
    DO 40 I = 1, M
        V(I) = 0
        JFIRST = (I+1)/2
        JLAST = MIN0(I,MH)
        DO 30 J = JFIRST, JLAST
            V(I) = V(I) + H(J)/(F(I-J+1)*F(2*J-I+1))
30  CONTINUE
        V(I) = ISIGN*V(I)
        ISIGN = -ISIGN
40  CONTINUE
    M = M
50  FA = 0
    A = .693147180559945309417232121458DO / 1
    DO 60 I = 1, M
        FA = FA + V(I)*P(I*A)
60  CONTINUE
    FA = A*FA
    RETURN
    END

```

C
C

```

SUBROUTINE EZWNR(ITABLE)
IMPLICIT REAL*8 (A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /INOUT/ TVAL(16),KVAL(7),SCLNT(10,16,3),RHOIN,
1ZSIN(3)
COMMON /LALAC/ V(50),M,N
COMMON /RHOZO/ RHO,Z,OSTART
EXTERNAL FWH

```

C
C

```

    SCALING WITH SIGSQB = 1
    RHO = RHOIN
    ALFSQB = (1-RHO)*(1-RHO)
    WTFCTR = DABS(1-RHO)
    Z = ZSIN(ITABLE)/WTFCTR

```

C

```

    DO 20 J=1,16
        T = TVAL(J)/ALFSQB
        CALL LINV(PWB,M,T,TERM3,V,M)
        EZWT = Z + (RHO-1.)*T + TERM3
        SCLNT(1,J,ITABLE) = EZWT*WTFCTR
20  CONTINUE

```

C

```

    RETURN
    END

```

C
C

```
DOUBLE PRECISION FUNCTION PWNR(S)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /RHOZO/ RHO,Z,OSTART
OMEGA = -(1-RHO)+ESQRT((1-RHO)*(1-RHO)+2*S)
PWNR = DEXP(-OMEGA*Z)/(S*OMEGA)
RETURN
END
```

C
C

```
SUBROUTINE RZMD1(ITABLE)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /INPUT/ TVAL(16),KVAL(7),SCLNT(10,16,3),RHOIN,
1ZSIN(3)
COMMON /LAPLAC/ V(50),B,N
COMMON /RHOZO/ RHO,Z,OSTART
EXTERNAL FMD1,DFMD1,DFMD1,DDFMD1
```

C
C

```
SCALING WITH SIGSQE = RHO
RHC = RHOIN
ALFSQE = (1-RHO)*(1-RHO)/RHO
WTFCTR = DABS(1-RHO)/RHO
OSTART = 1
Z = ZSIN(ITABLE)/WTFCTR
ELZ = DEXP(-RHO*Z)
```

C

```
IF (RHO .LE. 1.0) GO TO 10
CALL ZHOC(DFMD1,DDFMD1,1D-10,SBAR)
DELTA = -FMD1(SBAR)
START = SBAR + DELTA
CALL ZHOC(FMD1,DFMD1,START,OSTART)
10 CONTINUE
DO 20 J=1,16
T = TVAL(J)/ALFSQE
IF (T .LE. Z) EZIT = 0.
IF (T .GT. Z) CALL LINV(FMD1,B,T,HT,V,M)
IF (T .GT. Z) EZIT = HT + (T-Z)*ELZ
EZWT = Z + (RHO-1.)*T + EZIT
SCLNT(2,J,ITABLE) = EZWT*WTFCTR
20 CONTINUE
RETURN
END
```

C
C


```

DOUBLE PRECISION FUNCTION PBD1(S)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /RHOZO/ RHO,Z,OSTART
COMMON SS
EXTERNAL FOMD1,DFMD1
SS= S
CALL ZERO(FOMD1,DFMD1,OSTART,OMEGA)
OSTART = OMEGA
PBD1 = DEXP(-OMEGA*Z)/(S*OMEGA) - DEXP(-Z*(RHO+S))/(S*S)
RETURN
END

```

C
C

```

DOUBLE PRECISION FUNCTION PBD1(S)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /RHOZO/ RHO,Z,OSTART
PBD1 = S-RHO*(1-DEXP(-S))
RETURN
END

```

C
C

```

DOUBLE PRECISION FUNCTION DFMD1(S)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /RHOZO/ RHO,Z,OSTART
DFMD1 = 1-RHO*DEXP(-S)
RETURN
END

```

C
C

```

DOUBLE PRECISION FUNCTION DDFMD1(S)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /RHOZO/ RHO,Z,OSTART
DDFMD1 = RHO*DEXP(-S)
RETURN
END

```

C
C

```

DOUBLE PRECISION FUNCTION FOMD1(OMEGA)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON SS
FOMD1 = PBD1(OMEGA) - SS
RETURN
END

```

C

```

C      SUBROUTINE EZMEK(ITABLE)
        IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
        COMMON /IMOUT/ TVAL(16),KVAL(7),SCLWT(10,16,3),RHOIN,
12SIN(3)
        COMMON /LAPLAC/ V(50),M,N
        COMMON /RHOZOK/ RHO,Z,OSTART,K
        EXTERNAL FMEK,DFMEK,DDFMEK

C      SCALING WITH SIGSQK = (K+1)*RHO/K
C      RHO = RHOIN

C      DO 30 I=3,9
        K = KVAL(I-2)
        ALFSQK = K*(1-RHO)*(1-RHO)/((K+1)*RHO)
        WTFCTR = K*DABS(1-RHO)/((K+1)*RHO)
        OSTART = 1
        Z = ZSIN(ITABLE)/WTFCTR
        ELZ = DEXP(-RHO*Z)

C      IF (RHO .LE. 1.0) GO TO 10
        CALL ZERO(DFMEK,DDFMEK,0,SBAR)
        DELTA = -FMEK(SBAR)
        START = SBAR + DELTA
        CALL ZERO(FMEK,DFMEK,START,OSTART)
10     CONTINUE

C      DO 20 J=1,16
        T = TVAL(J)/ALFSQK
        IF (T .LE. Z) EZIT = 0.
        IF (T .GT. Z) CALL LINV(FMEK,M,T,HT,V,M)
        IF (T .GT. Z) EZIT = HT + (T-Z)*ELZ
        EZWT = Z + (RHO-1.)*T + EZIT
        SCLWT(I,J,ITABLE) = EZWT*WTFCTR
20     CONTINUE
30     CONTINUE
        RETURN
        END

C      DOUBLE PRECISION FUNCTION FMEK(S)
        IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
        COMMON /RHOZOK/ RHO,Z,OSTART,K
        COMMON SS
        EXTERNAL FMEK,DFMEK
        SS = S
        CALL ZERO(FMEK,DFMEK,OSTART,OMEGA)
        OSTART = OMEGA
        FMEK = DEXP(-OMEGA*Z)/(S*OMEGA) - DEXP(-Z*(RHO+S))/(S*S)
        RETURN
        END

```

C

```

DOUBLE PRECISION FUNCTION FMEK(S)
IMPLICIT REAL*8 (A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /RHOZOK/ RHO,2,CSTART,K
FMEK = S-RHC*(1-(K/(K+S))**K)
RETURN
END

```

C

C

```

DOUBLE PRECISION FUNCTION DFMEK(S)
IMPLICIT REAL*8 (A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /RHOZOK/ RHO,2,OSTART,K
DFMEK = 1-RHO*(K/(K+S))**(K+1)
RETURN
END

```

C

C

```

DOUBLE PRECISION FUNCTION DDFMEK(S)
IMPLICIT REAL*8 (A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /RHOZOK/ RHO,2,OSTART,K
DDFMEK = RHC*(K+1)*(K/(K+S))**(K+2)/K
RETURN
END

```

C

C

```

DOUBLE PRECISION FUNCTION FOMEK(OMEGA)
IMPLICIT REAL*8 (A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON SS
FOMEK = FMEK(OMEGA) - SS
RETURN
END

```

C

C

```

SUBROUTINE RZGAN(ITABLE)
IMPLICIT REAL*8 (A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /INOUT/ TVAL(16),KVAL(7),SCLWT(10,16,3),RHOIN,
1ZSIN(3)
COMMON /LAPLAC/ V(50),N,M
COMMON /RHOZO/ RHO,2,OSTART
EXTERNAL FGAM,FGAM,DFGAM

```

C

C

```

SCALING WITH SIGSQK = RHO
RHC = RHOIN
ALFSQA = (1-RHO)*(1-RHO)/RHO
WTFCTR = DABS(1-RHC)/RHO
OSTART = 1
Z = ZSIN(ITABLE)/WTFCTR

```

C


```

      IF (RHO .LE. 1.0) GO TO 10
      SBAR = RHO - 1
      DELTA = -FGAM(SBAR)
      START = SBAR + DELTA
      CALL ZERO(FGAM,DFGAM,START,OSTART)
10    CONTINUE
      DO 20 J=1,16
      T = TVAL(J)/ALFSUB
      IF (T .LE. Z) EZIT = 0.
      IF (T .GT. Z) CALL LINV(PGAM,N,T,EZIT,V,N)
      EZWT = Z + (RHO-1.)*T + EZIT
      SCLNT(10,J,ITABLE) = EZWT*WTFCTR
20    CONTINUE

      RETURN
      END

C
C
      DOUBLE PRECISION FUNCTION PGAM(S)
      IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
      COMMON /RHOZO/ RHO,Z,OSTART
      COMMON SS
      EXTERNAL FOGAM,DFGAM
      SS = S
      CALL ZERO(FOGAM,DFGAM,OSTART,OMEGA)
      OSTART = OMEGA
      PGAM = DEXP(-OMEGA*Z)/(S*OMEGA)
      RETURN
      END

C
C
      DOUBLE PRECISION FUNCTION FGAM(S)
      IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
      COMMON /RHOZO/ RHO,Z,OSTART
      FGAM = S-RHO*DLOG(1+S)
      RETURN
      END

C
C
      DOUBLE PRECISION FUNCTION DFGAM(S)
      IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
      COMMON /RHOZO/ RHO,Z,OSTART
      DFGAM = 1-RHO/(1+S)
      RETURN
      END

C
C

```

```

DOUBLE PRECISION FUNCTION FOGAM(OMEGA)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,N,M)
COMMON /RHOZO/ RHC,Z,OSTART
COMMON SS
FOGAM = FGAM(OMEGA) - SS
RETURN
END

```

C
C

```

SUBROUTINE ZERO(F,DF,XSTART,XZERO)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,N,M)
COMMON /NEWRAP/ TOLRNC,MAXITN
ITERAT = 0

```

C

```

X = XSTART
10 CONTINUE
XNEW = X - F(X)/DF(X)
RELACC = (XNEW-X)/XNEW
X = XNEW
ITERAT = ITERAT + 1
IF (ITERAT .GE. MAXITN) GO TO 50
IF (DABS(RELACC) .LE. TOLRNC) GO TO 20
GO TO 10
20 CONTINUE
XZERO = X
RETURN

```

C

```

50 WRITE (6,60)
60 FORMAT (1H1,'NEWTON-RAPHSON SEARCH IN SUBROUTINE ZERO',
1' EXCEEDED MAXIMUM ALLOWABLE NUMBER OF ITERATIONS.')
WRITE (6,62) MAXITN,RELACC,TOLRNC
62 FORMAT (1H0,'MAXITN =',I4,///,'RELACC =',D28.16,
1///,'TOLRNC =',D28.16)
STOP
END

```

C
C

```

SUBROUTINE ZOUT
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,N,M)
COMMON /INOUT/ TVAL(16),KVAL(7),SCLWT(10,16,3),RHOIN,
1ZSIN(3)
DIMENSION ISCLWT(10,16,3)

```

C

```

DO 90 NCOPY=1,5

```

C

```

WRITE(6,10)
10 FORMAT(1H1,///)
MARGIN = 4 + NCOPY
DO 12 IMGN=5,MARGIN
WRITE(6,11)
11 FORMAT(1H )
12 CONTINUE

```

```

C
DO 80 I=1,3
C
SCALOR = 1000.
IF (SCLMT(1,16,ITABLE) .GE. 10.) SCALOR = 100.
C
DO 14 I = 1,10
DO 14 J = 1,16
ISCLMT(I,J,ITABLE) = SCALOR*SCLMT(I,J,ITABLE) + 0.5
14 CONTINUE
C
IF (SCALOR .EQ. 100.) WRITE(6,16) RHOIN,ZSIN(ITABLE)
16 FORMAT(1H-,10X,'RHO =',F5.2,6X,'SCALED INITIAL SERVER',
1' LOAD =',F5.2,12X,'SCALED MEAN WAIT IN HUNDREDTHS')
C
IF (SCALOR .EQ. 1000.) WRITE(6,20) RHOIN,ZSIN(ITABLE)
20 FORMAT(1H-,10X,'RHO =',F5.2,6X,'SCALED INITIAL SERVER',
1' LOAD =',F5.2,11X,'SCALED MEAN WAIT IN THOUSANDTHS')
C
WRITE (6,25)
25 FORMAT(1H ,55X,'SCALED EPOCH')
C
WRITE (6,30) TVAL
30 FORMAT(1H ,21X,SF5.2,7F5.1)
C
WRITE (6,35)
35 FORMAT(1H ,21X,16(1X,'----'))
C
WRITE (6,40) (ISCLMT(1,J,ITABLE), J=1,16)
40 FORMAT(1H ,10X,'MEMBER ',16(1X,I4))
C
WRITE (6,45) (ISCLMT(2,J,ITABLE), J=1,16)
45 FORMAT(1H ,10X,'M1 QUEUE ',16(1X,I4))
C
DO 60 I=3,9
WRITE (6,50) KVAL(I-2), (ISCLMT(I,J,ITABLE), J=1,16)
50 FORMAT(1H ,10X,'M1 K=',F5.2,16(1X,I4))
60 CONTINUE
C
WRITE (6,70) (ISCLMT(10,J,ITABLE), J=1,16)
70 FORMAT(1H ,10X,'GAMMA INPUT',16(1X,I4))
C
60 CONTINUE
C
90 CONTINUE
C
RETURN
END

```


COMPUTER PROGRAM FOR APPENDIX TABLES, RHO EQUAL TO ONE

```

      IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
      COMMON /INOUT/ TVAL(16),KVAL(7),WAIT(10,10,3),ZIN(3)
      COMMON /LAPLAC/ V(50),M,N
      COMMON /NEWBAP/ TOLRNC,MAXITN

```

```

C      N=0
      N=34
      MAXITN=20
      TOLRNC=1.D-20

```

```

C      DO 60 IPAGE=1,3
      READ(5,20) (ZIN(1,ITABLE), ITABLE=1,3)
      READ(5,20) (TVAL(J), J=1,6)
      READ(5,20) (TVAL(J), J=7,12)
      READ(5,20) (TVAL(J), J=13,16)
20    FORMAT(6F10.0)

```

```

C      DO 50 ITABLE=1,3
      CALL RZWNK(ITABLE)
      CALL RZMD1(ITABLE)
      CALL RZBEK(ITABLE)
      CALL RZGAM(ITABLE)
50    CONTINUE
      CALL RZQUT
60    CONTINUE
      STOP
      END

```

```

C      C
      ELOCK DATA
      IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
      COMMON /INOUT/ TVAL(16),KVAL(7),WAIT(10,10,3),ZIN(3)
      DATA KVAL/4.,3.,2.,1.5,1.,.5,.2/
      END

```

```

C      C
      SUBROUTINE LINV(P,N,T,FA,V,M)
      IMPLICIT REAL*8(A-H,O-Z), INTEGER(1-N)
      COMMON/LAPLAC/EPOCH(10),APPROX(10,6)
      DIMENSION E(25), F(25), H(25), V(50)
      IF (M.EQ.N) GO TO 50
      NH = N/2
      E(1) = 2
      E(2) = 12
      F(1) = 1
      F(2) = 1
      DO 10 I = 2, NH
      E(I+1) = ( 2.D0*(2*I+1)*E(I) ) / I
      F(I+1) = F(I)*I
10    CONTINUE
      FNE = NH

```

```

      H(1) = 2/F(NH)
      DO 20 J = 2, NH
      H(J) = ((J**PWR)*E(J))/F(NH-J+1)
20    CONTINUE
      ISIGN = 2*(NH-2*(NH/2))-1
      DO 40 I = 1, N
      V(I) = 0
      JFIRST = (I+1)/2
      JLAST = MIN0(I,NH)
      DO 30 J = JFIRST, JLAST
      V(I) = V(I) + H(J)/(F(I-J+1)*F(2*J-I+1))
30    CONTINUE
      V(I) = ISIGN*V(I)
      ISIGN = -ISIGN
40    CONTINUE
      A = A
50    FA = 0
      A = .69314718055994530941723212145880 / T
      DO 60 I = 1, N
      FA = FA + V(I)*F(I*A)
60    CONTINUE
      FA = A*FA
      RETURN
      END

```

C
C

```

SUBROUTINE RZWRN(ITABLE)
  IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
  COMMON /INOUT/ TVAL(16),KVAL(7),WAIT(10,16,3),ZIN(3)
  COMMON /LAPLAC/ V(50),M,N
  COMMON /ZO/ Z,OSTART
  EXTERNAL PWR

```

C

```

      Z = ZIN(ITABLE)
      DO 20 J=1,16
      T = TVAL(J)
      CALL LINV(PWR,N,T,TERM3,V,M)
      WAIT(1,J,ITABLE) = Z + TERM3
20    CONTINUE
      RETURN
      END

```

C
C

```

DOUBLE PRECISION FUNCTION PWR(S)
  IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
  COMMON /ZO/ Z,OSTART
  OMEGA = DSQRT(2*S)
  PWR = DEXP(-OMEGA*Z)/(S*OMEGA)
  RETURN
  END

```

C
C

```

SUBROUTINE BZMD1(ITABLE)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /INOUT/ TVAL(16),KVAL(7),WAIT(10,16,3),ZIN(3)
COMMON /LAPLAC/ V(50),M,N
COMMON /ZO/ Z,OSTART
EXTERNAL FMD1,DFMD1
OSTART = 1
Z = ZIN(ITABLE)
ELZ = DEXP(-Z)

```

```

C
DO 20 J=1,16
T = TVAL(J)
IF (T.LE. Z) EZIT = 0.
IF (T.GT. Z) CALL LINV(FMD1,M,T,HT,V,M)
IF (T.GT. Z) EZIT = HT + (T-Z)*ELZ
20 WAIT(2,J,ITABLE) = Z + EZIT
CONTINUE
RETURN
END

```

```

C
C
DOUBLE PRECISION FUNCTION FMD1(S)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /ZO/ Z,OSTART
COMMON SS
EXTERNAL FOMD1,DFMD1
SS = S
CALL ZERO(FOMD1,DFMD1,OSTART,OMEGA)
OSTART = OMEGA
FMD1 = DEXP(-OMEGA*Z)/(S*OMEGA) - DEXP(-Z*(1+S))/(S*S)
RETURN
END

```

```

C
DOUBLE PRECISION FUNCTION FMD1(S)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /ZO/ Z,OSTART
FMD1 = S - 1. + DEXP(-S)
RETURN
END

```

```

C
DOUBLE PRECISION FUNCTION DFMD1(S)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /ZO/ Z,OSTART
DFMD1 = 1.-DEXP(-S)
RETURN
END

```

```

C
DOUBLE PRECISION FUNCTION FOMD1(OMEGA)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON SS
FOMD1 = FMD1(OMEGA) - SS
RETURN
END

```



```

SUBROUTINE RZMEK (ITABLE)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /INOUT/ TVAL(16),KVAL(7),WAIT(10,16,3),ZIN(3)
COMMON /LAPLAC/ V(50),A,N
COMMON /ZOK/ Z,OSTART,K
EXTERNAL PHEK,FMEK,DFMEK

```

C

```

DO 30 I=3,9
  K      = KVAL(I-2)
  OSTART = 1
  Z      = ZIN(ITABLE)
  ELZ    = DEXP(-Z*(K+1)/K)

```

C

```

DO 20 J=1,16
  T = TVAL(J)
  IF (T .LE. Z) EZIT = 0.
  IF (T .GT. Z) CALL LINV(PHEK,M,T,HT,V,M)
  IF (T .GT. Z) EZIT = HT + (T-Z)*ELZ
  WAIT(I,J,ITABLE) = Z + EZIT
20 CONTINUE
30 CONTINUE
RETURN
END

```

C

C

```

DOUBLE PRECISION FUNCTION PHEK(S)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /ZOK/ Z,OSTART,K
COMMON SS
EXTERNAL FOMEK,DFMEK
SS = S
CALL ZERC(FOMEK,DFMEK,OSTART,OMEGA)
OSTART = OMEGA
PHEK = DEXP(-OMEGA*Z)/(S*OMEGA) - DEXP(-Z*(S+(K+1)/K))/(S*S)
RETURN
END

```

C

C

```

DOUBLE PRECISION FUNCTION FMEK(S)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /ZOK/ Z,OSTART,K
FMEK = S-(K+1)*(1-((K+1)/(K+1+S))**K)/K
RETURN
END

```

C

C

```

DOUBLE PRECISION FUNCTION DFMEK(S)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /ZOK/ Z,OSTART,K
DFMEK = 1-((K+1)/(K+1+S))**(K+1)
RETURN
END

```

```

DOUBLE PRECISION FUNCTION FOMER(OMEGA)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON SS
FOMER = FMER(OMEGA) - SS
RETURN
END

```

C
C

```

SUBROUTINE RZGAM(ITABLE)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /INOUT/ TVAL(16), KVAL(7), WAIT(10,16,3), ZIN(3)
COMMON /LAPLAC/ V(50), M, N
COMMON /ZO/ Z, OSTART
EXTERNAL FGAM, FGGAM, DFGAM
OSTART = 1
Z = ZIN(ITABLE)

```

C

```

DO 20 J=1,16
T = TVAL(J)
IF (T.LE. 4) EZIT = 0.
IF (T.GT. 2) CALL LINV(PGAM,M,T,EZIT,V,M)
WAIT(10,J,ITABLE) = Z + EZIT
20 CONTINUE
RETURN
END

```

C
C

```

DOUBLE PRECISION FUNCTION FGAM(S)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /ZO/ Z, OSTART
COMMON SS
EXTERNAL FGGAM, DFGAM
SS = S
CALL ZERO(FOGAM,LFGAM,OSTART,OMEGA)
OSTART = OMEGA
FGAM = DEXP(-OMEGA*Z)/(S*OMEGA)
RETURN
END

```

C

```

DOUBLE PRECISION FUNCTION FGAM(S)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /ZO/ Z, OSTART
FGAM = S-DLOG(1+S)
RETURN
END

```

C

```

DOUBLE PRECISION FUNCTION DFGAM(S)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /ZO/ Z, OSTART
DFGAM = S/(1+S)
RETURN
END

```

```

C
DOUBLE PRECISION FUNCTION FOGAM(OMEGA)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON SS
FOGAZ = FGAH(OMEGA) - SS
RETURN
END

C
C
SUBROUTINE ZERO(F,DF,XSTART,XZERO)
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /NEWRAF/ TOLRNC,MAXITM
ITERAT = 0

C
X = XSTART
10 CONTINUE
XNEW = X - F(X)/DF(X)
RELACC = (XNEW-X)/XNEW
X = XNEW
ITERAT = ITERAT + 1
IF (ITERAT .GE. MAXITM) GO TO 50
IF (DABS(RELACC) .LE. TOLRNC) GO TO 20
GO TO 10
20 CONTINUE
XZERO = X
RETURN

C
50 WRITE (6,60)
60 FORMAT (1H1,'NEWTON-RAPHSON SEARCH IN SUBROUTINE ZERO',
1' EXCEEDED MAXIMUM ALLOWABLE NUMBER OF ITERATIONS.')
WRITE (6,62) MAXITM,RELACC,TOLRNC
62 FORMAT (1H0,'MAXITM =',I4,'/',,'RELACC =',D28.16,
1'/',,'TOLRNC =',D28.16)
STOP
END

C
C
SUBROUTINE SZOUT
IMPLICIT REAL*8(A-H,K,L,O-Z), INTEGER(I,J,M,N)
COMMON /INOUT/ TVAL(16),KVAL(7),WAIT(10,16,3),ZIN(3)
DIMENSION IWAIT(10,16,3)

C
DO 90 MCOPY=1,5

C
WRITE (6,10)
10 FORMAT (1H1,'/')
MARGIN = 6 + MCOPY
DO 12 IMAGE=7,MARGIN
WRITE (6,11)
11 FORMAT (1H )
12 CONTINUE

```



```

C      DO 60 I=1,3
C      SCALOR = 1000.
C      IF (WAIT(1,16,ITABLE) .GE. 10.) SCALOR = 100.
C
C      DO 14 J = 1,10
C      DO 14 J = 1,16
C      IWAIT(I,J,ITABLE) = SCALOR*WAIT(I,J,ITABLE) + 0.5
14  CONTINUE
C
C      WRITE(6,25)
25  FORMAT(1H-,10X,'PARAMETERS FOR THESE PROCESSES SELECTED',
1' SO THAT VAR(X(1))=1. AND E(X(1))=0.0, I.E., RHO=1.0')
C
C      IF (SCALOR .EQ. 100.) WRITE(6,16) ZIN(ITABLE)
16  FORMAT(1H-,10X,'MEAN WAIT IN HUNDREDTHS',
143X,'INITIAL SERVER LOAD = ',F3.1)
C
C      IF (SCALOR .EQ. 1000.) WRITE(6,20) ZIN(ITABLE)
20  FORMAT(1H-,10X,'MEAN WAIT IN THOUSANDTHS',
142X,'INITIAL SERVER LOAD = ',F3.1)
C
C      WRITE(6,30) TVAL
30  FORMAT(1H-,10X,'EPOCH:',5X,8F5.2,8F5.1)
C
C      WRITE(6,35)
35  FORMAT(1H-,21X,16(1X,'----'))
C
C      WRITE(6,40) (IWAIT(1,J,ITABLE), J=1,16)
40  FORMAT(1H-,10X,'MEANER ',16(1X,I4))
C
C      WRITE(6,45) (IWAIT(2,J,ITABLE), J=1,16)
45  FORMAT(1H-,10X,'ME1 QUEUE ',16(1X,I4))
C
C      DO 60 I=3,9
C      WRITE(6,50) KVAL(I-2), (IWAIT(I,J,ITABLE), J=1,16)
50  FORMAT(1H-,10X,'ME1 K=',F5.2,16(1X,I4))
60  CONTINUE
C
C      WRITE(6,70) (IWAIT(10,J,ITABLE), J=1,16)
70  FORMAT(1H-,10X,'GAMMA INPUT',16(1X,I4))
C
C      80  CONTINUE
C
C      90  CONTINUE
C
C      RETURN
C      END

```

APPENDIX C

TABLES OF SCALED EXPECTED SERVER LOAD

All tables for a specific value of ρ are grouped together.

The tables are arranged in the order shown in this list.

Each page contains results for three values of z^* .

ρ		z^*
2.0		
1.5	There are two pages of tables for each of these five values of ρ .	
1.4		1.0, 0.8, 0.6
1.3		0.4, 0.2, 0.0
1.2		
1.1		
1.0		
0.9	There are three pages of tables for each of these seven values of ρ .	4.0, 3.0, 2.0
0.8		1.0, 0.8, 0.6
0.7		0.4, 0.2, 0.0
0.6		
0.5		

RHO = 2.00

	0.63	0.80	1.00	1.20	1.40	1.60	1.80	2.00	SCALED	EPOCH	3.0	3.5	4.0	5.0	6.0	7.0	8.0
WIENER	1615	1824	2031	2237	2441	2645	2849	3052	---	3557	4060	4562	5064	5566	6067	6567	---
MDI QUEUE	1600	1800	2000	2203	2404	2605	2806	3007	---	3509	4010	4511	5012	5512	6013	6513	---
MEI K = 4.00	1600	1800	2000	2201	2402	2603	2804	3004	---	3506	4007	4508	5008	5508	6009	6509	---
MEI K = 3.00	1600	1800	2000	2201	2402	2602	2803	3003	---	3505	4006	4507	5007	5507	6008	6508	---
MEI K = 2.00	1600	1800	2000	2201	2401	2601	2802	3003	---	3504	4005	4506	5006	5506	6007	6507	---
MEI K = 1.50	1600	1800	2000	2200	2401	2601	2802	3003	---	3504	4004	4505	5005	5505	6006	6506	---
MEI K = 1.00	1600	1800	2000	2200	2400	2601	2802	3003	---	3503	4003	4504	5004	5504	6005	6505	---
MEI K = 0.50	1600	1800	2000	2200	2400	2600	2801	3001	---	3502	4002	4503	5003	5503	6004	6504	---
MEI K = 0.20	1600	1800	2000	2200	2400	2600	2800	3001	---	3501	4001	4502	5002	5502	6002	6502	---
GAMMA INPUT	1600	1800	2000	2200	2400	2600	2800	3000	---	3501	4001	4501	5001	5501	6001	6501	---

RHO = 2.00

	0.63	0.80	1.00	1.20	1.40	1.60	1.80	2.00	SCALED	EPOCH	3.0	3.5	4.0	5.0	6.0	7.0	8.0
WIENER	1433	1645	1855	2063	2269	2474	2678	2882	---	3388	3892	4395	4897	5399	5900	6400	---
MDI QUEUE	1400	1600	1806	2008	2211	2413	2615	2817	---	3319	3821	4322	4823	5324	5824	6324	---
MEI K = 4.00	1400	1600	1803	2005	2207	2409	2610	2811	---	3314	3815	4316	4816	5317	5817	6317	---
MEI K = 3.00	1400	1600	1802	2004	2206	2408	2609	2810	---	3312	3814	4315	4815	5316	5816	6316	---
MEI K = 2.00	1400	1600	1802	2003	2205	2406	2608	2809	---	3311	3812	4313	4813	5314	5814	6314	---
MEI K = 1.50	1400	1600	1801	2003	2204	2405	2607	2807	---	3309	3810	4311	4811	5312	5812	6312	---
MEI K = 1.00	1400	1600	1801	2002	2203	2404	2605	2806	---	3308	3809	4309	4810	5310	5810	6310	---
MEI K = 0.50	1400	1600	1800	2001	2202	2403	2604	2805	---	3305	3806	4307	4807	5307	5807	6307	---
MEI K = 0.20	1400	1600	1800	2001	2201	2402	2602	2803	---	3304	3804	4305	4805	5305	5805	6305	---
GAMMA INPUT	1400	1600	1800	2000	2201	2401	2601	2802	---	3302	3803	4303	4803	5303	5803	6303	---

RHO = 2.00

	0.63	0.80	1.00	1.20	1.40	1.60	1.80	2.00	SCALED	EPOCH	3.0	3.5	4.0	5.0	6.0	7.0	8.0
WIENER	1265	1482	1695	1905	2112	2319	2524	2728	---	3236	3741	4244	4746	5248	5749	6250	---
MDI QUEUE	1200	1413	1618	1823	2028	2231	2434	2636	---	3140	3642	4144	4645	5146	5646	6146	---
MEI K = 4.00	1200	1407	1612	1816	2020	2223	2425	2627	---	3130	3632	4133	4634	5135	5635	6135	---
MEI K = 3.00	1200	1406	1611	1815	2018	2221	2423	2625	---	3128	3630	4131	4632	5132	5632	6132	---
MEI K = 2.00	1200	1405	1609	1813	2016	2218	2420	2622	---	3125	3626	4127	4628	5129	5629	6129	---
MEI K = 1.50	1200	1404	1608	1811	2014	2216	2418	2620	---	3122	3624	4125	4625	5126	5626	6126	---
MEI K = 1.00	1200	1403	1606	1809	2011	2214	2415	2617	---	3119	3620	4121	4622	5122	5622	6122	---
MEI K = 0.50	1200	1402	1604	1806	2008	2210	2411	2612	---	3114	3616	4117	4617	5117	5617	6117	---
MEI K = 0.20	1200	1401	1602	1804	2006	2207	2408	2609	---	3111	3612	4112	4613	5113	5613	6113	---
GAMMA INPUT	1200	1400	1601	1803	2004	2205	2406	2607	---	3108	3609	4109	4609	5109	5610	6110	---

RHO = 2.00													
	SCALED INITIAL SERVER LOAD = 0.40					EPOCH	SCALED MEAN WAIT IN THOUSANDTHS						
	0.05	0.10	0.15	0.20	0.30		0.40	0.50	0.60	1.00	1.50	2.00	3.00
WIENER	454	518	583	647	772	1244	892	1008	1122	1559	2085	2599	3614
MDI QUEUE	450	500	550	600	700	1107	800	917	1028	1450	1966	2475	3483
ME1 K= 4.00	450	500	550	600	700	1091	800	911	1018	1439	1953	2460	3467
ME1 K= 3.00	450	500	550	600	700	1081	800	909	1016	1426	1950	2457	3463
ME1 K= 2.00	450	500	550	600	700	1076	800	908	1012	1432	1948	2452	3458
ME1 K= 1.50	450	500	550	600	700	1069	800	906	1010	1429	1941	2448	3453
ME1 K= 1.00	450	500	550	600	700	1059	800	905	1010	1425	1936	2442	3448
ME1 K= 0.50	450	500	550	600	700	1051	800	903	1007	1420	1929	2434	3439
ME1 K= 0.20	450	500	550	600	700	1043	800	902	1004	1415	1923	2428	3431
GAMMA INPUT	450	500	550	600	700	1207	800	901	1003	1411	1918	2422	3425

RHO = 2.00													
	SCALED INITIAL SERVER LOAD = 0.20					EPOCH	SCALED MEAN WAIT IN THOUSANDTHS						
	0.05	0.10	0.15	0.20	0.30		0.40	0.50	0.60	1.00	1.50	2.00	3.00
WIENER	283	369	447	520	655	1244	782	902	1019	1463	1992	2508	3523
MDI QUEUE	250	300	350	400	537	1107	662	779	889	1321	1841	2351	3360
ME1 K= 4.00	250	300	350	400	529	1091	648	762	870	1303	1821	2330	3338
ME1 K= 3.00	250	300	350	400	527	1087	645	759	865	1298	1816	2325	3333
ME1 K= 2.00	250	300	350	400	524	1081	641	754	865	1292	1808	2317	3324
ME1 K= 1.50	250	300	350	400	522	1076	638	750	861	1287	1803	2311	3318
ME1 K= 1.00	250	300	350	400	518	1069	633	745	855	1280	1795	2302	3309
ME1 K= 0.50	250	300	350	400	514	1059	627	738	846	1269	1782	2289	3295
ME1 K= 0.20	250	300	350	400	510	1051	622	731	839	1259	1771	2277	3282
GAMMA INPUT	250	300	350	400	508	1043	617	725	832	1250	1761	2267	3271

RHO = 2.00													
	SCALED INITIAL SERVER LOAD = 0.0					EPOCH	SCALED MEAN WAIT IN THOUSANDTHS						
	0.05	0.10	0.15	0.20	0.30		0.40	0.50	0.60	1.00	1.50	2.00	3.00
WIENER	205	306	392	469	609	1205	738	860	978	1425	1955	2472	3488
MDI QUEUE	95	182	263	338	475	1051	600	716	829	1266	1787	2298	3308
ME1 K= 4.00	94	179	256	327	459	1030	581	697	811	1243	1763	2273	3282
ME1 K= 3.00	94	178	254	324	455	1025	577	693	806	1238	1757	2267	3275
ME1 K= 2.00	93	176	251	320	449	1017	570	686	799	1230	1748	2257	3265
ME1 K= 1.50	93	174	248	317	445	1011	566	681	793	1223	1741	2250	3257
ME1 K= 1.00	92	172	244	312	439	1002	559	674	785	1214	1731	2239	3246
ME1 K= 0.50	92	168	239	305	430	989	548	662	773	1199	1715	2232	3229
ME1 K= 0.20	90	164	233	299	422	976	539	651	761	1186	1700	2207	3212
GAMMA INPUT	86	160	228	293	414	964	530	641	751	1173	1686	2192	3197

RHO = 1.50									
SCALED INITIAL SERVER LOAD = 1.00									
	0.30	0.40	0.50	0.60	0.70	0.80	1.00	SCALED EPOCH	
WIENER	1304	1408	1512	1616	1720	1824	2031	2237	1.40
MDI QUEUE	1300	1400	1500	1601	1703	1804	2007	2210	1.40
ME1 K = 4.00	1300	1400	1500	1601	1702	1803	2005	2208	1.40
ME1 K = 3.00	1300	1400	1500	1601	1701	1802	2005	2207	1.40
ME1 K = 2.00	1300	1400	1500	1600	1701	1802	2004	2206	1.40
ME1 K = 1.50	1300	1400	1500	1600	1701	1801	2004	2206	1.40
ME1 K = 1.00	1300	1400	1500	1600	1701	1801	2003	2205	1.40
ME1 K = 0.50	1300	1400	1500	1600	1700	1801	2002	2204	1.40
ME1 K = 0.20	1300	1400	1500	1600	1700	1800	2001	2203	1.40
GAMMA INPUT	1300	1400	1500	1600	1700	1800	2001	2202	1.40
SCALED MEAN WAIT IN THOUSANDTHS									
	1.6	1.8	2.0	2.5	3.0	5.0	8.0		
WIENER	2645	2849	3052	3557	4060	6066	9067		
MDI QUEUE	2615	2817	3019	3522	4024	6027	9028		
ME1 K = 4.00	2612	2814	3015	3518	4019	6022	9023		
ME1 K = 3.00	2611	2813	3014	3517	4018	6021	9021		
ME1 K = 2.00	2610	2811	3013	3515	4017	6019	9020		
ME1 K = 1.50	2609	2811	3012	3514	4016	6018	9018		
ME1 K = 1.00	2608	2809	3010	3513	4014	6016	9017		
ME1 K = 0.50	2607	2808	3008	3510	4011	6013	9014		
ME1 K = 0.20	2605	2806	3007	3508	4009	6011	9011		
GAMMA INPUT	2604	2805	3005	3507	4008	6009	9009		

RHO = 1.50									
SCALED INITIAL SERVER LOAD = 0.80									
	0.30	0.40	0.50	0.60	0.70	0.80	1.00	SCALED EPOCH	
WIENER	1112	1219	1327	1433	1540	1645	1855	2063	1.40
MDI QUEUE	1100	1200	1303	1406	1510	1613	1818	2023	1.40
ME1 K = 4.00	1100	1200	1302	1404	1507	1609	1814	2018	1.40
ME1 K = 3.00	1100	1200	1302	1404	1506	1608	1813	2017	1.40
ME1 K = 2.00	1100	1200	1301	1403	1505	1607	1812	2016	1.40
ME1 K = 1.50	1100	1200	1301	1403	1505	1607	1811	2015	1.40
ME1 K = 1.00	1100	1200	1301	1402	1504	1606	1809	2013	1.40
ME1 K = 0.50	1100	1200	1300	1401	1503	1604	1807	2010	1.40
ME1 K = 0.20	1100	1200	1300	1401	1502	1603	1806	2008	1.40
GAMMA INPUT	1100	1200	1300	1400	1501	1602	1804	2007	1.40
SCALED MEAN WAIT IN THOUSANDTHS									
	1.6	1.8	2.0	2.5	3.0	5.0	8.0		
WIENER	2474	2678	2882	3388	3892	5899	8901		
MDI QUEUE	2430	2633	2835	3339	3842	5846	8847		
ME1 K = 4.00	2425	2627	2829	3333	3835	5839	8839		
ME1 K = 3.00	2424	2626	2828	3331	3834	5837	8838		
ME1 K = 2.00	2422	2624	2826	3329	3831	5834	8835		
ME1 K = 1.50	2420	2622	2824	3327	3829	5832	8833		
ME1 K = 1.00	2418	2620	2822	3325	3827	5830	8830		
ME1 K = 0.50	2415	2617	2818	3321	3823	5825	8826		
ME1 K = 0.20	2412	2614	2815	3318	3819	5822	8822		
GAMMA INPUT	2410	2612	2813	3315	3816	5818	8819		

RHO = 1.50									
SCALED INITIAL SERVER LOAD = 0.60									
	0.30	0.40	0.50	0.60	0.70	0.80	1.00	SCALED EPOCH	
WIENER	930	1044	1155	1265	1374	1482	1695	1905	1.40
MDI QUEUE	900	1009	1115	1221	1327	1432	1641	1848	1.40
ME1 K = 4.00	900	1005	1111	1216	1322	1426	1634	1841	1.40
ME1 K = 3.00	900	1005	1110	1215	1320	1425	1633	1839	1.40
ME1 K = 2.00	900	1004	1109	1214	1318	1423	1632	1838	1.40
ME1 K = 1.50	900	1003	1108	1212	1317	1421	1628	1834	1.40
ME1 K = 1.00	900	1002	1106	1211	1315	1419	1626	1831	1.40
ME1 K = 0.50	900	1002	1105	1208	1312	1416	1622	1827	1.40
ME1 K = 0.20	900	1001	1103	1206	1310	1413	1618	1823	1.40
GAMMA INPUT	900	1000	1102	1205	1308	1410	1615	1819	1.40
SCALED MEAN WAIT IN THOUSANDTHS									
	1.6	1.8	2.0	2.5	3.0	5.0	8.0		
WIENER	2319	2524	2728	3236	3741	5748	8750		
MDI QUEUE	2258	2462	2665	3170	3673	5678	8679		
ME1 K = 4.00	2250	2453	2656	3160	3663	5668	8669		
ME1 K = 3.00	2248	2451	2654	3158	3661	5665	8666		
ME1 K = 2.00	2245	2448	2650	3155	3657	5661	8662		
ME1 K = 1.50	2242	2445	2648	3152	3655	5658	8659		
ME1 K = 1.00	2239	2442	2644	3148	3651	5654	8655		
ME1 K = 0.50	2236	2439	2639	3142	3645	5648	8648		
ME1 K = 0.20	2234	2437	2634	3137	3639	5642	8642		
GAMMA INPUT	2229	2432	2629	3132	3634	5637	8637		

RHO = 1.50									
SCALED INITIAL SERVER LOAD = 0.40									
0.03	0.10	0.15	0.20	0.30	0.40	0.50	0.60	EPOCH	SCALED
0.03	0.10	0.15	0.20	0.30	0.40	0.50	0.60	0.80	1.00
1.0	1.5	2.0	3.0	4.0	6.0	8.0	10.0	12.0	14.0
WIENER	454	518	583	647	712	772	835	892	947
MDI QUEUE	450	500	550	600	650	700	750	800	850
ME1 K= 4.00	450	500	550	600	650	700	750	800	850
ME1 K= 3.00	450	500	550	600	650	700	750	800	850
ME1 K= 2.00	450	500	550	600	650	700	750	800	850
ME1 K= 1.50	450	500	550	600	650	700	750	800	850
ME1 K= 1.00	450	500	550	600	650	700	750	800	850
ME1 K= 0.50	450	500	550	600	650	700	750	800	850
ME1 K= 0.20	450	500	550	600	650	700	750	800	850
GAMMA INPUT	450	500	550	600	650	700	750	800	850

RHO = 1.50									
SCALED INITIAL SERVER LOAD = 0.20									
0.03	0.10	0.15	0.20	0.30	0.40	0.50	0.60	EPOCH	SCALED
0.03	0.10	0.15	0.20	0.30	0.40	0.50	0.60	0.80	1.00
1.0	1.5	2.0	3.0	4.0	6.0	8.0	10.0	12.0	14.0
WIENER	283	369	447	520	595	672	749	826	903
MDI QUEUE	250	300	350	400	450	500	550	600	650
ME1 K= 4.00	250	300	350	400	450	500	550	600	650
ME1 K= 3.00	250	300	350	400	450	500	550	600	650
ME1 K= 2.00	250	300	350	400	450	500	550	600	650
ME1 K= 1.50	250	300	350	400	450	500	550	600	650
ME1 K= 1.00	250	300	350	400	450	500	550	600	650
ME1 K= 0.50	250	300	350	400	450	500	550	600	650
ME1 K= 0.20	250	300	350	400	450	500	550	600	650
GAMMA INPUT	250	300	350	400	450	500	550	600	650

RHO = 1.50									
SCALED INITIAL SERVER LOAD = 0.0									
0.03	0.10	0.15	0.20	0.30	0.40	0.50	0.60	EPOCH	SCALED
0.03	0.10	0.15	0.20	0.30	0.40	0.50	0.60	0.80	1.00
1.0	1.5	2.0	3.0	4.0	6.0	8.0	10.0	12.0	14.0
WIENER	205	306	392	469	546	623	700	777	854
MDI QUEUE	131	232	315	387	459	531	603	675	747
ME1 K= 4.00	127	222	303	376	449	521	594	667	739
ME1 K= 3.00	125	220	301	374	447	519	592	665	737
ME1 K= 2.00	124	217	297	369	442	514	587	659	732
ME1 K= 1.50	122	214	294	366	438	510	583	655	728
ME1 K= 1.00	120	211	289	361	433	505	578	650	723
ME1 K= 0.50	117	206	283	354	426	498	570	643	715
ME1 K= 0.20	114	201	277	347	419	491	563	635	707
GAMMA INPUT	111	196	271	341	411	483	555	627	700

RHO = 1.40																
WIENER																
MDI QUEUE																
MEI K= 4.00																
MEI K= 3.00																
MEI K= 2.00																
MEI K= 1.50																
MEI K= 1.00																
MEI K= 0.50																
MEI K= 0.20																
GAMMA INPUT																
0.20	0.30	0.40	0.50	0.60	0.70	0.80	0.90	SCALED	EPOCH	1.5	2.0	2.5	3.0	4.0	6.0	8.0
1201	1304	1408	1512	1616	1720	1824	1928	1824	2031	2544	3052	3557	4060	5064	7067	9067
1200	1300	1400	1501	1602	1704	1806	1908	1806	2010	2518	3023	3526	4028	5031	7032	9033
1200	1300	1400	1501	1602	1703	1805	1906	1805	2008	2514	3019	3522	4024	5026	7027	9028
1200	1300	1400	1500	1601	1703	1804	1906	1804	2007	2514	3018	3521	4023	5025	7026	9027
1200	1300	1400	1500	1601	1702	1804	1905	1804	2006	2513	3017	3520	4021	5023	7025	9025
1200	1300	1400	1500	1601	1702	1803	1905	1803	2005	2512	3016	3518	4020	5022	7023	9024
1200	1300	1400	1500	1601	1702	1803	1904	1803	2005	2510	3014	3517	4018	5020	7021	9022
1200	1300	1400	1500	1600	1701	1802	1903	1802	2004	2509	3012	3514	4016	5018	7019	9019
1200	1300	1400	1500	1600	1701	1801	1902	1801	2003	2507	3010	3512	4014	5015	7016	9016
1200	1300	1400	1500	1600	1700	1801	1902	1801	2002	2506	3009	3510	4012	5013	7014	9014

RHO = 1.40																
WIENER																
MDI QUEUE																
MEI K= 4.00																
MEI K= 3.00																
MEI K= 2.00																
MEI K= 1.50																
MEI K= 1.00																
MEI K= 0.50																
MEI K= 0.20																
GAMMA INPUT																
0.20	0.30	0.40	0.50	0.60	0.70	0.80	0.90	SCALED	EPOCH	1.5	2.0	2.5	3.0	4.0	6.0	8.0
1005	1112	1219	1327	1433	1540	1645	1750	1645	1855	2372	2882	3388	3892	4897	6900	8901
1000	1100	1202	1306	1409	1513	1617	1720	1617	1823	2334	2841	3346	3848	4851	6853	8854
1000	1100	1201	1304	1407	1510	1613	1716	1613	1819	2329	2836	3340	3842	4845	6847	8847
1000	1100	1201	1304	1406	1509	1612	1715	1612	1818	2328	2834	3338	3841	4843	6845	8845
1000	1100	1201	1303	1406	1508	1611	1714	1611	1816	2326	2832	3336	3838	4841	6842	8843
1000	1100	1201	1302	1404	1507	1609	1712	1609	1814	2325	2831	3334	3836	4839	6840	8841
1000	1100	1200	1301	1403	1505	1607	1709	1607	1811	2320	2828	3332	3834	4836	6838	8838
1000	1100	1200	1301	1402	1504	1606	1708	1606	1810	2317	2825	3328	3830	4832	6833	8834
1000	1100	1200	1301	1402	1503	1605	1706	1605	1808	2314	2822	3324	3826	4828	6829	8830
1000	1100	1200	1301	1402	1503	1605	1706	1605	1808	2314	2819	3321	3823	4825	6826	8826

RHO = 1.40																
WIENER																
MDI QUEUE																
MEI K= 4.00																
MEI K= 3.00																
MEI K= 2.00																
MEI K= 1.50																
MEI K= 1.00																
MEI K= 0.50																
MEI K= 0.20																
GAMMA INPUT																
0.20	0.30	0.40	0.50	0.60	0.70	0.80	0.90	SCALED	EPOCH	1.5	2.0	2.5	3.0	4.0	6.0	8.0
815	930	1044	1155	1265	1374	1482	1589	1482	1695	2216	2728	3236	3741	4746	6749	8750
800	905	1013	1120	1227	1333	1439	1544	1439	1648	2164	2673	3179	3682	4686	6688	8689
800	903	1009	1116	1222	1328	1433	1538	1433	1642	2157	2665	3170	3673	4677	6679	8679
800	903	1009	1115	1221	1327	1432	1536	1432	1640	2155	2663	3168	3671	4675	6677	8677
800	902	1007	1113	1219	1325	1430	1534	1430	1638	2152	2660	3165	3668	4671	6673	8673
800	902	1007	1112	1218	1323	1428	1532	1428	1636	2150	2658	3162	3665	4668	6670	8671
800	901	1006	1111	1216	1321	1426	1530	1426	1633	2147	2654	3159	3661	4664	6666	8666
800	901	1004	1109	1213	1318	1422	1526	1422	1629	2142	2649	3153	3656	4658	6660	8660
800	900	1003	1107	1211	1315	1419	1522	1419	1626	2137	2644	3148	3650	4652	6654	8654
800	900	1002	1105	1209	1313	1416	1519	1416	1622	2133	2639	3143	3645	4647	6648	8649

RHO = 1.40																
	SCALED INITIAL SERVER LOAD = 0.40					EPOCH	SCALED MEAN WAIT IN THOUSANDTHS									
	0.05	0.10	0.15	0.20	0.30		0.40	0.50	0.60	1.0	1.5	2.0	3.0	4.0	6.0	8.0
WIENER	454	518	583	647	772	892	1008	1122	1343	1559	2085	2599	3614	4619	6623	8624
MD1 QUEUE	450	500	550	611	727	843	956	1067	1284	1497	2017	2529	3539	4543	6546	8547
ME1 K= 4.00	450	500	550	607	722	836	948	1059	1275	1487	2007	2517	3527	4531	6533	8534
ME1 K= 3.00	450	500	550	606	721	835	947	1057	1273	1485	2004	2514	3524	4528	6530	8531
ME1 K= 2.00	450	500	550	605	719	832	944	1054	1270	1481	2000	2510	3520	4523	6525	8526
ME1 K= 1.50	450	500	550	604	718	830	942	1052	1267	1478	1997	2507	3516	4519	6522	8522
ME1 K= 1.00	450	500	550	603	716	828	939	1048	1263	1474	1992	2502	3511	4514	6516	8516
ME1 K= 0.50	450	500	550	602	713	824	934	1043	1257	1468	1985	2494	3502	4505	6507	8507
ME1 K= 0.20	450	500	550	601	710	820	930	1038	1252	1462	1978	2486	3494	4497	6499	8499
GAMMA INPUT	450	500	550	601	708	817	926	1034	1246	1456	1971	2479	3487	4489	6491	8491

RHO = 1.40																
	SCALED INITIAL SERVER LOAD = 0.20					EPOCH	SCALED MEAN WAIT IN THOUSANDTHS									
	0.05	0.10	0.15	0.20	0.30		0.40	0.50	0.60	1.0	1.5	2.0	3.0	4.0	6.0	8.0
WIENER	283	369	447	520	655	782	902	1019	1244	1453	1992	2508	3523	4530	6534	8535
MD1 QUEUE	250	317	394	461	592	715	833	948	1169	1385	1909	2422	3434	4438	6441	8442
ME1 K= 4.00	250	313	385	454	583	705	822	936	1157	1372	1895	2407	3419	4423	6426	8426
ME1 K= 3.00	250	312	384	452	581	703	820	933	1154	1369	1892	2404	3415	4419	6422	8422
ME1 K= 2.00	250	310	381	449	577	699	815	929	1150	1364	1886	2398	3409	4413	6415	8416
ME1 K= 1.50	250	309	379	447	574	696	812	926	1146	1360	1882	2394	3404	4408	6411	8411
ME1 K= 1.00	250	307	377	443	571	691	807	921	1140	1354	1876	2387	3397	4401	6403	8404
ME1 K= 0.50	250	305	373	438	564	684	800	913	1132	1345	1866	2377	3386	4390	6392	8392
ME1 K= 0.20	250	304	369	434	559	678	793	905	1123	1336	1856	2366	3376	4379	6381	8381
GAMMA INPUT	250	303	366	430	553	672	786	898	1116	1328	1847	2357	3365	4369	6370	8371

RHO = 1.40																
	SCALED INITIAL SERVER LOAD = 0.0					EPOCH	SCALED MEAN WAIT IN THOUSANDTHS									
	0.05	0.10	0.15	0.20	0.30		0.40	0.50	1.0	1.5	2.0	3.0	4.0	6.0	8.0	
WIENER	205	306	392	469	609	738	860	978	1205	1425	1955	2472	3488	4494	6499	8500
MD1 QUEUE	143	244	325	401	527	663	783	899	1123	1339	1865	2378	3391	4396	6399	8392
ME1 K= 4.00	136	234	316	390	525	651	770	886	1109	1325	1849	2362	3374	4378	6381	8389
ME1 K= 3.00	136	232	314	388	523	648	767	883	1105	1321	1846	2358	3370	4374	6377	8377
ME1 K= 2.00	134	229	310	384	518	643	762	877	1100	1316	1839	2347	3363	4367	6370	8371
ME1 K= 1.50	133	227	307	380	515	639	758	873	1095	1311	1835	2342	3358	4362	6365	8365
ME1 K= 1.00	130	223	303	376	510	634	753	867	1089	1304	1827	2339	3350	4354	6357	8357
ME1 K= 0.50	127	218	297	369	502	625	743	858	1079	1294	1816	2327	3338	4341	6344	8344
ME1 K= 0.20	123	213	291	362	494	617	735	849	1069	1283	1805	2316	3325	4329	6331	8331
GAMMA INPUT	120	209	286	356	487	609	726	840	1060	1274	1794	2305	3314	4317	6319	8319

RHO = 1.30													
SCALED INITIAL SERVER LOAD = 1.00													
	0.15	0.20	0.25	0.30	0.40	0.50	0.60	0.70	SCALED	EPOCH	0.9	1.0	1.5
WIENER	1150	1201	1252	1304	1402	1512	1616	1720	1824	1928	2031	2134	2244
MD1 QUEUE	1150	1200	1250	1300	1401	1503	1605	1707	1809	1911	2013	2116	2222
ME1 K= 4.00	1150	1200	1250	1300	1401	1502	1603	1705	1807	1909	2011	2113	2219
ME1 K= 3.00	1150	1200	1250	1300	1400	1502	1603	1705	1807	1909	2011	2113	2219
ME1 K= 2.00	1150	1200	1250	1300	1400	1501	1603	1704	1806	1908	2010	2112	2218
ME1 K= 1.50	1150	1200	1250	1300	1400	1501	1602	1704	1806	1908	2010	2112	2218
ME1 K= 1.00	1150	1200	1250	1300	1400	1501	1602	1704	1806	1908	2010	2112	2218
ME1 K= 0.50	1150	1200	1250	1300	1400	1501	1602	1704	1806	1908	2010	2112	2218
ME1 K= 0.20	1150	1200	1250	1300	1400	1500	1601	1702	1803	1905	2006	2107	2210
GAMMA INPUT	1150	1200	1250	1300	1400	1500	1601	1702	1803	1904	2005	2106	2209
WAIT IN THOUSANDTHS													
	0.15	0.20	0.25	0.30	0.40	0.50	0.60	0.70	0.80	0.9	1.0	1.5	2.0
WIENER	4.00	5.06	6.06	7.06	8.06	9.06	10.06	11.06	12.06	13.06	14.06	15.06	16.06
MD1 QUEUE	4.00	5.06	6.06	7.06	8.06	9.06	10.06	11.06	12.06	13.06	14.06	15.06	16.06
ME1 K= 4.00	4.00	5.06	6.06	7.06	8.06	9.06	10.06	11.06	12.06	13.06	14.06	15.06	16.06
ME1 K= 3.00	4.00	5.06	6.06	7.06	8.06	9.06	10.06	11.06	12.06	13.06	14.06	15.06	16.06
ME1 K= 2.00	4.00	5.06	6.06	7.06	8.06	9.06	10.06	11.06	12.06	13.06	14.06	15.06	16.06
ME1 K= 1.50	4.00	5.06	6.06	7.06	8.06	9.06	10.06	11.06	12.06	13.06	14.06	15.06	16.06
ME1 K= 1.00	4.00	5.06	6.06	7.06	8.06	9.06	10.06	11.06	12.06	13.06	14.06	15.06	16.06
ME1 K= 0.50	4.00	5.06	6.06	7.06	8.06	9.06	10.06	11.06	12.06	13.06	14.06	15.06	16.06
ME1 K= 0.20	4.00	5.06	6.06	7.06	8.06	9.06	10.06	11.06	12.06	13.06	14.06	15.06	16.06
GAMMA INPUT	4.00	5.06	6.06	7.06	8.06	9.06	10.06	11.06	12.06	13.06	14.06	15.06	16.06

RHO = 1.30													
SCALED INITIAL SERVER LOAD = 0.80													
	0.15	0.20	0.25	0.30	0.40	0.50	0.60	0.70	SCALED	EPOCH	0.9	1.0	1.5
WIENER	950	1005	1058	1112	1219	1327	1433	1540	1645	1750	1855	1959	2064
MD1 QUEUE	950	1000	1050	1101	1205	1309	1413	1518	1621	1725	1828	1931	2034
ME1 K= 4.00	950	1000	1050	1101	1203	1307	1411	1515	1618	1722	1825	1928	2031
ME1 K= 3.00	950	1000	1050	1101	1203	1307	1411	1515	1618	1722	1825	1928	2031
ME1 K= 2.00	950	1000	1050	1101	1203	1307	1411	1515	1618	1722	1825	1928	2031
ME1 K= 1.50	950	1000	1050	1101	1203	1307	1411	1515	1618	1722	1825	1928	2031
ME1 K= 1.00	950	1000	1050	1101	1203	1307	1411	1515	1618	1722	1825	1928	2031
ME1 K= 0.50	950	1000	1050	1101	1203	1307	1411	1515	1618	1722	1825	1928	2031
ME1 K= 0.20	950	1000	1050	1101	1203	1307	1411	1515	1618	1722	1825	1928	2031
GAMMA INPUT	950	1000	1050	1101	1203	1307	1411	1515	1618	1722	1825	1928	2031
WAIT IN THOUSANDTHS													
	0.15	0.20	0.25	0.30	0.40	0.50	0.60	0.70	0.80	0.9	1.0	1.5	2.0
WIENER	6.00	7.00	8.00	9.00	10.00	11.00	12.00	13.00	14.00	15.00	16.00	17.00	18.00
MD1 QUEUE	6.00	7.00	8.00	9.00	10.00	11.00	12.00	13.00	14.00	15.00	16.00	17.00	18.00
ME1 K= 4.00	6.00	7.00	8.00	9.00	10.00	11.00	12.00	13.00	14.00	15.00	16.00	17.00	18.00
ME1 K= 3.00	6.00	7.00	8.00	9.00	10.00	11.00	12.00	13.00	14.00	15.00	16.00	17.00	18.00
ME1 K= 2.00	6.00	7.00	8.00	9.00	10.00	11.00	12.00	13.00	14.00	15.00	16.00	17.00	18.00
ME1 K= 1.50	6.00	7.00	8.00	9.00	10.00	11.00	12.00	13.00	14.00	15.00	16.00	17.00	18.00
ME1 K= 1.00	6.00	7.00	8.00	9.00	10.00	11.00	12.00	13.00	14.00	15.00	16.00	17.00	18.00
ME1 K= 0.50	6.00	7.00	8.00	9.00	10.00	11.00	12.00	13.00	14.00	15.00	16.00	17.00	18.00
ME1 K= 0.20	6.00	7.00	8.00	9.00	10.00	11.00	12.00	13.00	14.00	15.00	16.00	17.00	18.00
GAMMA INPUT	6.00	7.00	8.00	9.00	10.00	11.00	12.00	13.00	14.00	15.00	16.00	17.00	18.00

RHO = 1.30													
SCALED INITIAL SERVER LOAD = 0.60													
	0.15	0.20	0.25	0.30	0.40	0.50	0.60	0.70	SCALED	EPOCH	0.9	1.0	1.5
WIENER	750	816	873	930	1044	1155	1265	1374	1482	1589	1695	1799	1904
MD1 QUEUE	750	802	855	909	1018	1126	1234	1341	1447	1552	1657	1761	1866
ME1 K= 4.00	750	801	853	907	1015	1122	1229	1336	1442	1547	1651	1755	1860
ME1 K= 3.00	750	801	853	906	1014	1121	1228	1335	1440	1545	1650	1754	1859
ME1 K= 2.00	750	800	853	906	1013	1120	1227	1333	1438	1543	1648	1752	1857
ME1 K= 1.50	750	800	852	905	1012	1119	1225	1331	1436	1541	1646	1750	1855
ME1 K= 1.00	750	800	852	904	1011	1117	1224	1329	1434	1539	1644	1748	1853
ME1 K= 0.50	750	800	851	902	1009	1115	1221	1326	1431	1536	1641	1746	1851
ME1 K= 0.20	750	800	851	902	1007	1113	1218	1323	1428	1533	1638	1743	1848
GAMMA INPUT	750	800	850	902	1006	1111	1216	1321	1426	1531	1636	1741	1846
WAIT IN THOUSANDTHS													
	0.15	0.20	0.25	0.30	0.40	0.50	0.60	0.70	0.80	0.9	1.0	1.5	2.0
WIENER	4.00	5.06	6.06	7.06	8.06	9.06	10.06	11.06	12.06	13.06	14.06	15.06	16.06
MD1 QUEUE	4.00	5.06	6.06	7.06	8.06	9.06	10.06	11.06	12.06	13.06	14.06	15.06	16.06
ME1 K= 4.00	4.00	5.06	6.06	7.06	8.06	9.06	10.06	11.06	12.06	13.06	14.06	15.06	16.06
ME1 K= 3.00	4.00	5.06	6.06	7.06	8.06	9.06	10.06	11.06	12.06	13.06	14.06	15.06	16.06
ME1 K= 2.00	4.00	5.06	6.06	7.06	8.06	9.06	10.06	11.06	12.06	13.06	14.06	15.06	16.06
ME1 K= 1.50	4.00	5.06	6.06	7.06	8.06	9.06	10.06	11.06	12.06	13.06	14.06	15.06	16.06
ME1 K= 1.00	4.00	5.06	6.06	7.06	8.06	9.06	10.06	11.06	12.06	13.06	14.06	15.06	16.06
ME1 K= 0.50	4.00	5.06	6.06	7.06	8.06	9.06	10.06	11.06	12.06	13.06	14.06	15.06	16.06
ME1 K= 0.20	4.00	5.06	6.06	7.06	8.06	9.06	10.06	11.06	12.06	13.06	14.06	15.06	16.06
GAMMA INPUT	4.00	5.06	6.06	7.06	8.06	9.06	10.06	11.06	12.06	13.06	14.06	15.06	16.06

RHO = 1.30											
SCALED INITIAL SERVER LOAD = 0.40											
	0.05	0.10	0.15	0.20	0.30	0.40	0.50	0.60	EPOCH	SCALED MEAN	WAIT IN THOUSANDTHS
WIENER	454	518	583	647	772	892	1008	1122	1343	1559	1772
MD1 QUEUE	450	500	558	617	736	852	966	1077	1296	1519	1736
ME1 K= 4.00	450	500	555	613	731	846	959	1070	1288	1501	1718
ME1 K= 3.00	450	500	554	612	730	845	958	1066	1283	1496	1713
ME1 K= 2.00	450	500	553	611	728	843	955	1064	1281	1493	1709
ME1 K= 1.50	450	500	553	610	726	841	953	1064	1281	1493	1709
ME1 K= 1.00	450	500	552	609	724	838	950	1061	1272	1483	1709
ME1 K= 0.50	450	500	551	605	718	831	942	1051	1266	1478	1709
GAMMA INPUT	450	500	551	604	716	828	938	1047	1262	1472	1709

RHO = 1.30											
SCALED INITIAL SERVER LOAD = 0.20											
	0.05	0.10	0.15	0.20	0.30	0.40	0.50	0.60	EPOCH	SCALED MEAN	WAIT IN THOUSANDTHS
WIENER	283	369	447	520	655	782	902	1019	1244	1463	1682
MD1 QUEUE	250	331	403	473	605	729	847	962	1184	1401	1618
ME1 K= 4.00	250	324	397	466	597	720	838	952	1174	1390	1607
ME1 K= 3.00	250	322	395	465	595	718	836	950	1172	1387	1603
ME1 K= 2.00	250	321	393	462	592	714	832	946	1168	1383	1603
ME1 K= 1.50	250	319	391	460	589	712	829	943	1164	1380	1603
ME1 K= 1.00	250	317	389	457	586	708	825	939	1160	1375	1603
ME1 K= 0.50	250	314	385	452	580	701	818	932	1152	1367	1603
GAMMA INPUT	250	312	381	448	575	696	812	925	1145	1359	1603

RHO = 1.30											
SCALED INITIAL SERVER LOAD = 0.0											
	0.05	0.10	0.15	0.20	0.30	0.40	0.50	0.60	EPOCH	SCALED MEAN	WAIT IN THOUSANDTHS
WIENER	205	306	392	469	609	738	860	978	1205	1425	1645
MD1 QUEUE	158	256	339	414	551	678	799	915	1139	1357	1575
ME1 K= 4.00	149	248	331	405	542	668	788	904	1128	1345	1562
ME1 K= 3.00	145	247	329	403	539	666	785	901	1125	1342	1558
ME1 K= 2.00	145	244	326	400	536	661	781	897	1120	1337	1552
ME1 K= 1.50	143	242	323	397	533	658	778	894	1117	1333	1548
ME1 K= 1.00	139	238	319	393	528	654	773	888	1111	1328	1543
ME1 K= 0.50	136	229	314	387	521	646	765	880	1103	1318	1535
GAMMA INPUT	133	224	303	375	508	632	750	865	1086	1301	1524

RHO = 1.20

	0.10	0.15	0.20	0.25	0.30	0.35	0.40	0.50	EPOCH	0.8	1.0	2.0	3.0	4.0	6.0	8.0
WIENER	1100	1150	1201	1252	1304	1356	1408	1512	1616	1824	2031	3052	4060	5064	7067	9067
MD1 QUEUE	1100	1150	1200	1250	1301	1351	1402	1505	1607	1813	2018	3034	4041	5044	7046	9046
ME1 K= 4.00	1100	1150	1200	1250	1300	1351	1402	1504	1606	1811	2016	3031	4037	5040	7042	9043
ME1 K= 3.00	1100	1150	1200	1250	1300	1351	1402	1504	1606	1811	2015	3030	4037	5039	7041	9042
ME1 K= 2.00	1100	1150	1200	1250	1300	1351	1401	1503	1605	1810	2014	3028	4034	5037	7039	9039
ME1 K= 1.50	1100	1150	1200	1250	1300	1351	1401	1503	1605	1809	2013	3027	4033	5036	7037	9038
ME1 K= 1.00	1100	1150	1200	1250	1300	1351	1401	1502	1604	1808	2012	3025	4031	5033	7035	9035
ME1 K= 0.50	1100	1150	1200	1250	1300	1350	1401	1502	1603	1807	2011	3023	4029	5031	7032	9033
ME1 K= 0.20	1100	1150	1200	1250	1300	1350	1400	1501	1603	1806	2010	3021	4027	5029	7030	9031
GAMMA INPUT	1100	1150	1200	1250	1300	1350	1400	1501	1603	1806	2010	3021	4027	5029	7030	9031

RHO = 1.20

	0.10	0.15	0.20	0.25	0.30	0.35	0.40	0.50	EPOCH	0.8	1.0	2.0	3.0	4.0	6.0	8.0
WIENER	900	952	1005	1058	1112	1165	1219	1327	1433	1645	1855	2882	3892	4897	6900	8901
MD1 QUEUE	900	950	1000	1052	1104	1156	1208	1313	1418	1628	1835	2857	3866	4870	6872	8872
ME1 K= 4.00	900	950	1000	1051	1103	1155	1207	1311	1416	1625	1832	2853	3861	4865	6867	8868
ME1 K= 3.00	900	950	1000	1051	1102	1154	1206	1310	1415	1624	1831	2852	3860	4864	6866	8867
ME1 K= 2.00	900	950	1000	1051	1102	1154	1206	1310	1415	1623	1830	2851	3859	4862	6864	8865
ME1 K= 1.50	900	950	1000	1051	1102	1154	1206	1310	1414	1622	1829	2849	3857	4861	6863	8863
ME1 K= 1.00	900	950	1000	1051	1102	1153	1205	1309	1413	1621	1828	2848	3855	4858	6860	8861
ME1 K= 0.50	900	950	1000	1050	1101	1153	1204	1308	1412	1619	1826	2845	3852	4855	6857	8857
ME1 K= 0.20	900	950	1000	1050	1101	1152	1203	1307	1410	1618	1824	2842	3849	4852	6854	8854
GAMMA INPUT	900	950	1000	1050	1101	1152	1203	1306	1409	1616	1822	2839	3846	4849	6850	8851

RHO = 1.20

	0.10	0.15	0.20	0.25	0.30	0.35	0.40	0.50	EPOCH	0.8	1.0	2.0	3.0	4.0	6.0	8.0
WIENER	703	759	816	873	930	987	1044	1155	1265	1482	1695	2728	3741	4746	6749	8750
MD1 QUEUE	700	751	805	859	914	970	1025	1134	1242	1456	1667	2696	3706	4711	6713	8714
ME1 K= 4.00	700	751	804	858	912	967	1021	1130	1239	1452	1663	2690	3700	4704	6707	8708
ME1 K= 3.00	700	751	803	857	911	965	1020	1129	1236	1449	1660	2687	3696	4703	6705	8706
ME1 K= 2.00	700	751	803	857	911	965	1020	1129	1236	1449	1660	2687	3696	4700	6703	8703
ME1 K= 1.50	700	750	803	856	910	965	1019	1128	1235	1448	1658	2685	3695	4698	6701	8701
ME1 K= 1.00	700	750	802	856	909	964	1018	1126	1234	1446	1656	2682	3692	4696	6698	8699
ME1 K= 0.50	700	750	802	855	908	962	1016	1124	1231	1443	1653	2678	3687	4691	6693	8694
ME1 K= 0.20	700	750	801	854	907	961	1014	1122	1229	1440	1650	2674	3683	4686	6689	8689
GAMMA INPUT	700	750	801	853	906	959	1013	1120	1226	1438	1646	2670	3679	4682	6684	8685

RHO = 1.20

	0.02	0.04	0.06	0.08	0.10	0.15	0.20	0.30	EPOCH	0.6	0.8	1.0	2.0	4.0	6.0	8.0
WIENER	420	442	466	492	518	583	647	772	892	1122	1343	1559	2599	4619	6623	8624
MDI QUEUE	420	440	460	480	504	564	625	746	863	1090	1309	1523	2558	4575	6578	8579
MEI K= 4.00	420	440	460	480	502	561	622	742	858	1084	1303	1517	2551	4567	6570	8571
MEI K= 3.00	420	440	460	480	502	560	620	739	855	1081	1299	1515	2549	4565	6568	8569
MEI K= 2.00	420	440	460	480	501	559	619	738	854	1079	1297	1513	2546	4562	6565	8566
MEI K= 1.50	420	440	460	480	501	557	617	736	852	1077	1294	1511	2544	4560	6563	8563
MEI K= 1.00	420	440	460	480	501	557	615	733	848	1073	1290	1503	2535	4550	6553	8553
MEI K= 0.50	420	440	460	480	500	556	614	730	845	1069	1286	1498	2529	4544	6547	8547
MEI K= 0.20	420	440	460	480	500	555	612	728	842	1065	1282	1494	2524	4538	6541	8541
GAMMA INPUT																

RHO = 1.20

	0.02	0.04	0.06	0.08	0.10	0.15	0.20	0.30	EPOCH	0.6	0.8	1.0	2.0	4.0	6.0	8.0
WIENER	228	265	301	336	369	447	520	655	782	1019	1244	1463	2508	4530	6534	8535
MDI QUEUE	220	240	277	308	340	416	487	619	744	978	1202	1418	2459	4477	6481	8482
MEI K= 4.00	220	240	273	305	336	410	481	612	737	971	1194	1410	2450	4468	6471	8472
MEI K= 3.00	220	240	272	304	335	408	478	609	733	969	1192	1408	2447	4465	6469	8469
MEI K= 2.00	220	240	271	302	333	408	478	609	733	966	1189	1405	2444	4462	6465	8465
MEI K= 1.50	220	240	270	301	332	406	476	607	731	964	1186	1403	2441	4459	6462	8462
MEI K= 1.00	220	240	269	300	330	404	473	604	728	960	1183	1399	2437	4454	6457	8458
MEI K= 0.50	220	240	267	297	327	400	469	600	723	955	1177	1393	2430	4446	6449	8450
MEI K= 0.20	220	240	266	295	325	397	466	595	718	950	1171	1386	2423	4439	6442	8443
GAMMA INPUT																

RHO = 1.20

	0.02	0.04	0.06	0.08	0.10	0.15	0.20	0.30	EPOCH	0.6	0.8	1.0	2.0	4.0	6.0	8.0
WIENER	123	181	227	269	306	392	469	609	738	978	1205	1425	2472	4494	6499	8500
MDI QUEUE	91	147	193	233	271	354	430	568	695	933	1158	1376	2419	4438	6442	8442
MEI K= 4.00	87	142	188	228	265	348	423	561	688	925	1150	1367	2409	4428	6431	8432
MEI K= 3.00	86	141	186	226	263	346	422	559	686	923	1148	1365	2406	4425	6429	8429
MEI K= 2.00	85	139	184	224	261	344	419	556	683	920	1144	1362	2402	4421	6424	8425
MEI K= 1.50	83	138	183	223	259	342	417	554	680	917	1141	1359	2399	4418	6421	8422
MEI K= 1.00	82	136	180	220	256	339	414	550	677	913	1137	1355	2395	4413	6416	8416
MEI K= 0.50	80	133	177	216	252	334	409	545	671	907	1131	1348	2387	4404	6407	8408
MEI K= 0.20	78	130	173	212	248	330	404	540	665	901	1124	1341	2379	4396	6399	8400
GAMMA INPUT																

RHO = 1.10									
SCALED INITIAL SERVER LOAD = 4.00									
0.20	0.30	0.40	0.50	0.60	0.70	0.80	0.90	SCALED	EPOCH
420	430	440	450	460	470	480	490	500	1.00
WIENER	430	440	450	460	470	480	490	500	1.00
MDI QUEUE	430	440	450	460	470	480	490	500	1.00
MEI K = 4.00	430	440	450	460	470	480	490	500	1.00
MEI K = 3.00	430	440	450	460	470	480	490	500	1.00
MEI K = 2.00	430	440	450	460	470	480	490	500	1.00
MEI K = 1.50	430	440	450	460	470	480	490	500	1.00
MEI K = 1.00	430	440	450	460	470	480	490	500	1.00
MEI K = 0.50	430	440	450	460	470	480	490	500	1.00
MEI K = 0.20	430	440	450	460	470	480	490	500	1.00
GAMMA INPUT	430	440	450	460	470	480	490	500	1.00
SCALED MEAN WAIT IN HUNDRETHS									
1.5	2.0	3.0	4.0	6.0	10.0	20.0			
550	600	700	800	1000	1400	2400			
550	600	700	800	1000	1400	2400			
550	600	700	800	1000	1400	2400			
550	600	700	800	1000	1400	2400			
550	600	700	800	1000	1400	2400			
550	600	700	800	1000	1400	2400			
550	600	700	800	1000	1400	2400			
550	600	700	800	1000	1400	2400			
550	600	700	800	1000	1400	2400			
550	600	700	800	1000	1400	2400			

RHO = 1.10									
SCALED INITIAL SERVER LOAD = 3.00									
0.20	0.30	0.40	0.50	0.60	0.70	0.80	0.90	SCALED	EPOCH
320	330	340	350	360	370	380	390	400	1.00
WIENER	330	340	350	360	370	380	390	400	1.00
MDI QUEUE	330	340	350	360	370	380	390	400	1.00
MEI K = 4.00	330	340	350	360	370	380	390	400	1.00
MEI K = 3.00	330	340	350	360	370	380	390	400	1.00
MEI K = 2.00	330	340	350	360	370	380	390	400	1.00
MEI K = 1.50	330	340	350	360	370	380	390	400	1.00
MEI K = 1.00	330	340	350	360	370	380	390	400	1.00
MEI K = 0.50	330	340	350	360	370	380	390	400	1.00
MEI K = 0.20	330	340	350	360	370	380	390	400	1.00
GAMMA INPUT	330	340	350	360	370	380	390	400	1.00
SCALED MEAN WAIT IN HUNDRETHS									
1.5	2.0	3.0	4.0	6.0	10.0	20.0			
450	500	600	700	900	1300	2300			
450	500	600	700	900	1300	2300			
450	500	600	700	900	1300	2300			
450	500	600	700	900	1300	2300			
450	500	600	700	900	1300	2300			
450	500	600	700	900	1300	2300			
450	500	600	700	900	1300	2300			
450	500	600	700	900	1300	2300			
450	500	600	700	900	1300	2300			
450	500	600	700	900	1300	2300			
450	500	600	700	900	1300	2300			

RHO = 1.10									
SCALED INITIAL SERVER LOAD = 2.00									
0.20	0.30	0.40	0.50	0.60	0.70	0.80	0.90	SCALED	EPOCH
220	230	240	250	260	270	280	290	300	1.00
WIENER	230	240	250	260	270	280	290	300	1.00
MDI QUEUE	230	240	250	260	270	280	290	300	1.00
MEI K = 4.00	230	240	250	260	270	280	290	300	1.00
MEI K = 3.00	230	240	250	260	270	280	290	300	1.00
MEI K = 2.00	230	240	250	260	270	280	290	300	1.00
MEI K = 1.50	230	240	250	260	270	280	290	300	1.00
MEI K = 1.00	230	240	250	260	270	280	290	300	1.00
MEI K = 0.50	230	240	250	260	270	280	290	300	1.00
MEI K = 0.20	230	240	250	260	270	280	290	300	1.00
GAMMA INPUT	230	240	250	260	270	280	290	300	1.00
SCALED MEAN WAIT IN HUNDRETHS									
1.5	2.0	3.0	4.0	6.0	10.0	20.0			
350	400	500	600	800	1200	2200			
350	400	500	600	800	1200	2200			
350	400	500	600	800	1200	2200			
350	400	500	600	800	1200	2200			
350	400	500	600	800	1200	2200			
350	400	500	600	800	1200	2200			
350	400	500	600	800	1200	2200			
350	400	500	600	800	1200	2200			
350	400	500	600	800	1200	2200			
350	400	500	600	800	1200	2200			
350	400	500	600	800	1200	2200			

RHO = 1.10													
SCALED INITIAL SERVER LOAD = 1.00													
	0.05	0.10	0.15	0.20	0.30	0.40	0.50	0.60	SCALED	EPOCH	SCALED	MEAN	WAIT
	0.05	0.10	0.15	0.20	0.30	0.40	0.50	0.60	0.80	1.00	1.50	2.00	3.00
WIENER	1050	1100	1150	1201	1304	1408	1512	1616	1824	2031	2544	3052	4060
MD1 QUEUE	1050	1100	1150	1200	1302	1404	1508	1611	1818	2023	2535	3042	4049
ME1 K= 4.00	1050	1100	1150	1200	1302	1404	1507	1610	1817	2022	2533	3040	4047
ME1 K= 3.00	1050	1100	1150	1200	1301	1404	1507	1610	1816	2022	2533	3039	4047
ME1 K= 2.00	1050	1100	1150	1200	1301	1404	1507	1610	1816	2021	2532	3039	4046
ME1 K= 1.50	1050	1100	1150	1200	1301	1403	1506	1609	1815	2021	2531	3038	4045
ME1 K= 1.00	1050	1100	1150	1200	1301	1403	1506	1608	1815	2020	2531	3037	4044
ME1 K= 0.50	1050	1100	1150	1200	1301	1403	1506	1608	1814	2019	2529	3036	4043
ME1 K= 0.20	1050	1100	1150	1200	1301	1403	1505	1608	1813	2018	2528	3035	4041
GAMMA INPUT	1050	1100	1150	1200	1301	1402	1505	1607	1813	2018	2527	3033	4040
	0.05	0.10	0.15	0.20	0.30	0.40	0.50	0.60	0.80	1.00	1.50	2.00	3.00
	0.05	0.10	0.15	0.20	0.30	0.40	0.50	0.60	0.80	1.00	1.50	2.00	3.00
WIENER	850	900	952	1005	1112	1219	1327	1433	1645	1855	2372	2882	3892
MD1 QUEUE	850	900	951	1002	1107	1213	1319	1425	1635	1844	2359	2868	3878
ME1 K= 4.00	850	900	950	1002	1106	1212	1318	1423	1634	1842	2357	2866	3875
ME1 K= 3.00	850	900	950	1002	1106	1212	1317	1422	1633	1841	2355	2865	3874
ME1 K= 2.00	850	900	950	1001	1106	1211	1317	1422	1632	1840	2355	2864	3873
ME1 K= 1.50	850	900	950	1001	1105	1211	1317	1422	1632	1840	2355	2863	3872
ME1 K= 1.00	850	900	950	1001	1105	1210	1316	1421	1631	1839	2353	2862	3871
ME1 K= 0.50	850	900	950	1001	1105	1210	1315	1420	1630	1838	2352	2860	3869
ME1 K= 0.20	850	900	950	1001	1104	1209	1314	1419	1628	1836	2350	2858	3867
GAMMA INPUT	850	900	950	1001	1104	1208	1313	1418	1627	1835	2348	2857	3865

RHO = 1.10													
SCALED INITIAL SERVER LOAD = 0.80													
	0.05	0.10	0.15	0.20	0.30	0.40	0.50	0.60	SCALED	EPOCH	SCALED	MEAN	WAIT
	0.05	0.10	0.15	0.20	0.30	0.40	0.50	0.60	0.80	1.00	1.50	2.00	3.00
WIENER	850	900	952	1005	1112	1219	1327	1433	1645	1855	2372	2882	3892
MD1 QUEUE	850	900	951	1002	1107	1213	1319	1425	1635	1844	2359	2868	3878
ME1 K= 4.00	850	900	950	1002	1106	1212	1318	1423	1634	1842	2357	2866	3875
ME1 K= 3.00	850	900	950	1002	1106	1212	1317	1422	1633	1841	2355	2865	3874
ME1 K= 2.00	850	900	950	1001	1106	1211	1317	1422	1632	1840	2355	2864	3873
ME1 K= 1.50	850	900	950	1001	1105	1211	1317	1422	1632	1840	2355	2863	3872
ME1 K= 1.00	850	900	950	1001	1105	1210	1316	1421	1631	1839	2353	2862	3871
ME1 K= 0.50	850	900	950	1001	1105	1210	1315	1420	1630	1838	2352	2860	3869
ME1 K= 0.20	850	900	950	1001	1104	1209	1314	1419	1628	1836	2350	2858	3867
GAMMA INPUT	850	900	950	1001	1104	1208	1313	1418	1627	1835	2348	2857	3865

RHO = 1.10													
SCALED INITIAL SERVER LOAD = 0.60													
	0.05	0.10	0.15	0.20	0.30	0.40	0.50	0.60	SCALED	EPOCH	SCALED	MEAN	WAIT
	0.05	0.10	0.15	0.20	0.30	0.40	0.50	0.60	0.80	1.00	1.50	2.00	3.00
WIENER	650	703	759	816	930	1044	1155	1265	1482	1695	2216	2728	3741
MD1 QUEUE	650	701	754	809	921	1033	1143	1253	1468	1679	2199	2710	3722
ME1 K= 4.00	650	701	754	809	920	1031	1141	1250	1465	1677	2196	2707	3718
ME1 K= 3.00	650	701	754	808	920	1031	1141	1250	1465	1676	2195	2706	3717
ME1 K= 2.00	650	700	753	808	919	1029	1139	1248	1464	1675	2194	2705	3716
ME1 K= 1.50	650	700	753	808	918	1028	1138	1247	1463	1674	2193	2704	3715
ME1 K= 1.00	650	700	753	807	918	1028	1138	1247	1461	1673	2191	2702	3713
ME1 K= 0.50	650	700	752	806	917	1027	1137	1245	1459	1671	2189	2700	3710
ME1 K= 0.20	650	700	752	806	916	1026	1135	1244	1458	1668	2187	2697	3708
GAMMA INPUT	650	700	752	805	915	1024	1134	1242	1456	1666	2184	2695	3705

RHO = 1.10

	0.02	0.04	0.06	0.08	0.10	0.15	0.20	0.40	SCALED	EPOCH	0.6	1.0	2.0	3.0	4.0	6.0	8.0
WIENER	420	442	466	492	518	563	647	892	1343	1122	1325	1559	2599	3614	4619	6623	8624
MD1 QUEUE	420	440	462	485	510	572	635	876	1325	1104	1321	1540	2577	3590	4595	6599	8600
ME1 K= 4.00	420	440	461	484	508	571	633	873	1320	1100	1320	1535	2572	3585	4590	6593	8594
ME1 K= 3.00	420	440	461	484	508	569	631	871	1319	1098	1318	1534	2570	3583	4586	6590	8592
ME1 K= 2.00	420	440	461	483	507	568	630	869	1316	1097	1316	1531	2567	3579	4584	6588	8588
ME1 K= 1.50	420	440	460	483	506	567	628	867	1313	1092	1311	1526	2563	3576	4581	6584	8585
ME1 K= 1.00	420	440	460	482	505	566	627	865	1311	1092	1311	1525	2560	3572	4577	6580	8581
ME1 K= 0.50	420	440	460	482	505	564	625	863	1308	1089	1308	1522	2557	3569	4574	6577	8577
GAMMA INPUT	420	440	460	482	505	564	625	863	1308	1089	1308	1522	2557	3569	4574	6577	8577

RHO = 1.10

	0.02	0.04	0.06	0.08	0.10	0.15	0.20	0.40	SCALED	EPOCH	0.8	1.0	2.0	3.0	4.0	6.0	8.0
WIENER	220	265	301	336	369	447	520	782	1244	1019	1221	1453	2508	3523	4530	6534	8535
MD1 QUEUE	220	253	287	321	354	430	502	761	1221	997	1217	1439	2481	3496	4501	6505	8506
ME1 K= 4.00	220	251	285	318	351	427	499	757	1216	993	1216	1433	2476	3489	4495	6498	8499
ME1 K= 3.00	220	250	284	317	349	425	497	755	1214	990	1214	1431	2473	3487	4492	6496	8497
ME1 K= 2.00	220	250	283	316	348	424	496	754	1213	989	1213	1430	2471	3485	4491	6494	8495
ME1 K= 1.50	220	249	282	315	347	423	494	752	1210	987	1210	1426	2469	3483	4488	6492	8492
ME1 K= 1.00	220	248	280	313	345	421	492	749	1207	983	1207	1424	2465	3478	4484	6487	8488
ME1 K= 0.50	220	247	279	311	343	418	489	746	1204	980	1204	1420	2461	3474	4479	6483	8484
ME1 K= 0.20	220	246	278	310	341	416	487	743	1200	977	1200	1417	2457	3470	4475	6479	8479
GAMMA INPUT	220	246	278	310	341	416	487	743	1200	977	1200	1417	2457	3470	4475	6479	8479

RHO = 1.10

	0.02	0.04	0.06	0.08	0.10	0.15	0.20	0.40	SCALED	EPOCH	0.8	1.0	2.0	3.0	4.0	6.0	8.0
WIENER	123	181	227	269	306	392	469	738	1205	978	1205	1425	2472	3488	4494	6499	8500
MD1 QUEUE	106	163	209	250	287	372	448	715	1180	954	1180	1399	2443	3458	4464	6468	8469
ME1 K= 4.00	103	160	206	246	284	368	444	711	1175	949	1175	1394	2438	3452	4458	6462	8463
ME1 K= 3.00	101	158	204	244	281	365	441	708	1172	946	1172	1390	2434	3449	4455	6458	8459
ME1 K= 2.00	101	157	203	243	280	364	440	706	1171	945	1171	1389	2432	3447	4453	6456	8457
ME1 K= 1.50	99	155	201	242	279	363	438	704	1168	943	1168	1386	2430	3444	4450	6453	8454
ME1 K= 1.00	97	153	199	239	276	360	435	701	1164	939	1164	1382	2425	3439	4445	6449	8449
ME1 K= 0.50	95	151	196	236	273	357	432	698	1161	935	1161	1378	2421	3435	4440	6444	8445
ME1 K= 0.20	94	149	194	234	271	354	429	694	1157	932	1157	1375	2417	3430	4436	6439	8440
GAMMA INPUT	94	149	194	234	271	354	429	694	1157	932	1157	1375	2417	3430	4436	6439	8440

PARAMETERS FOR THESE PROCESSES SELECTED SO THAT VAR(X(1))=1. AND E(X(1))=0.0. I.E., RHO=1.0.																
MEAN WAIT IN THOUSANDTHS																
EPOCH:	2.00	3.00	4.00	5.00	6.00	7.00	8.00	9.00	10.00	15.00	20.00	25.00	30.00	40.00	SER VER LOAD =	
WVENER	2101	2213	2333	2454	2572	2687	2799	2907	3012	3493	3919	4304	4658	5296	6385	7314
WMQ1 QUEUE	2000	2092	2200	2311	2424	2536	2644	2751	2854	3332	3756	4140	4494	5131	6219	7148
WMQ1 K= 4.00	2000	2077	2178	2287	2398	2509	2617	2722	2825	3302	3725	4109	4462	5099	6187	7116
WMQ1 K= 3.00	2000	2073	2173	2281	2392	2502	2610	2716	2818	3294	3718	4101	4454	5091	6179	7108
WMQ1 K= 2.00	2000	2067	2158	2262	2382	2493	2599	2704	2807	3282	3705	4088	4441	5078	6166	7094
WMQ1 K= 1.50	2000	2062	2154	2265	2374	2483	2590	2695	2798	3273	3695	4078	4431	5068	6155	7084
WMQ1 K= 1.00	2000	2056	2149	2254	2362	2471	2577	2682	2784	3258	3680	4063	4416	5052	6140	7069
WMQ1 K= 0.50	2000	2046	2135	2237	2346	2451	2557	2660	2762	3234	3656	4038	4391	5027	6113	7041
WMQ1 K= 0.20	2000	2038	2129	2221	2326	2431	2536	2639	2740	3211	3632	4014	4366	5001	6088	7015
CAMMA INPUT	2000	2031	2109	2206	2308	2413	2517	2619	2720	3188	3608	3990	4344	4976	6062	6989

PARAMETERS FOR THESE PROCESSES SELECTED SO THAT VAR(X(1))=1. AND E(X(1))=0.0. I.E., RHO=1.0
 MEAN WAIT IN THOUSANDS
 EPOCH: 0.63 0.80 1.00 1.50 2.00 3.00 4.00 5.00 6.0 8.0 10.0 15.0 20.0 40.0 60.0 80.0

WIENER	1072	1116	1167	1286	1399	1606	1791	1960	2115	2296	2648	3193	3657	5109	6232	7181
MD1 QUEUE	1000	1000	1000	1146	1238	1444	1629	1756	1951	2231	2483	3027	3491	4943	6065	7014
ME1 K= 4.00	1000	1000	1000	1115	1215	1419	1600	1767	1921	2201	2452	2995	3459	4911	6033	6982
ME1 K= 3.00	1000	1000	1000	1109	1213	1412	1593	1760	1914	2194	2445	2988	3452	4903	6025	6974
ME1 K= 2.00	1000	1000	1000	1101	1204	1401	1582	1748	1902	2181	2432	2975	3439	4890	6012	6961
ME1 K= 1.50	1000	1000	1000	1095	1197	1393	1573	1739	1893	2172	2422	2965	3428	4879	6001	6950
ME1 K= 1.00	1000	1000	1000	1087	1187	1381	1561	1726	1879	2158	2408	2950	3413	4863	5985	6934
ME1 K= 0.50	1000	1000	1000	1075	1171	1362	1540	1704	1857	2134	2384	2925	3388	4838	5959	6907
ME1 K= 0.20	1000	1000	1000	1064	1156	1344	1520	1683	1835	2112	2361	2901	3364	4812	5933	6881
GAMMA INPUT	1000	1000	1000	1055	1143	1327	1501	1663	1814	2090	2338	2878	3339	4787	5907	6855

PARAMETERS FOR THESE PROCESSES SELECTED SO THAT VAR(X(1))=1. AND E(X(1))=0.0. I.E., RHO=1.0
 MEAN WAIT IN THOUSANDS
 EPOCH: 0.63 0.80 1.00 1.50 2.00 3.00 4.00 5.00 6.0 8.0 10.0 15.0 20.0 40.0 60.0 80.0

WIENER	921	981	1040	1179	1304	1527	1722	1897	2058	2346	2603	3156	3625	5087	6213	7165
MD1 QUEUE	800	800	882	1026	1140	1362	1557	1732	1892	2180	2437	2989	3459	4920	6047	6998
ME1 K= 4.00	800	800	865	996	1117	1335	1528	1702	1862	2150	2407	2958	3427	4888	6014	6966
ME1 K= 3.00	800	800	861	990	1111	1328	1521	1695	1855	2143	2399	2951	3419	4880	6006	6958
ME1 K= 2.00	800	800	855	981	1101	1317	1510	1684	1843	2130	2387	2938	3407	4867	5993	6944
ME1 K= 1.50	800	800	850	974	1093	1309	1501	1674	1834	2121	2377	2928	3396	4856	5982	6934
ME1 K= 1.00	800	800	844	964	1082	1296	1488	1661	1820	2106	2362	2913	3381	4841	5967	6918
ME1 K= 0.50	800	800	835	949	1064	1276	1466	1639	1797	2083	2338	2888	3356	4815	5940	6891
ME1 K= 0.20	800	800	828	935	1048	1257	1446	1617	1775	2060	2315	2864	3331	4789	5914	6865
GAMMA INPUT	800	800	821	923	1032	1240	1427	1597	1754	2038	2292	2840	3307	4764	5889	6839

PARAMETERS FOR THESE PROCESSES SELECTED SO THAT VAR(X(1))=1. AND E(X(1))=0.0. I.E., RHO=1.0
 MEAN WAIT IN THOUSANDS
 EPOCH: 0.63 0.80 1.00 1.50 2.00 3.00 4.00 5.00 6.0 8.0 10.0 15.0 20.0 40.0 60.0 80.0

WIENER	795	868	937	1092	1228	1464	1667	1848	2013	2307	2568	3127	3600	5069	6199	7153
MD1 QUEUE	600	700	781	925	1066	1299	1501	1682	1846	2141	2402	2961	3434	4902	6032	6986
ME1 K= 4.00	600	684	753	902	1037	1270	1472	1652	1816	2110	2371	2929	3402	4870	6000	6953
ME1 K= 3.00	600	680	747	896	1030	1263	1465	1645	1809	2103	2364	2922	3394	4862	5992	6945
ME1 K= 2.00	600	673	738	887	1020	1252	1453	1633	1797	2091	2351	2909	3381	4849	5979	6932
ME1 K= 1.50	600	668	732	879	1012	1243	1444	1624	1788	2081	2341	2899	3371	4839	5968	6921
ME1 K= 1.00	600	662	723	868	1000	1230	1431	1610	1774	2067	2327	2884	3356	4823	5952	6905
ME1 K= 0.50	600	652	710	852	981	1210	1409	1588	1751	2043	2303	2859	3331	4797	5926	6879
ME1 K= 0.20	600	644	698	835	964	1191	1389	1566	1729	2020	2279	2835	3306	4771	5900	6853
GAMMA INPUT	600	636	688	822	948	1172	1369	1545	1708	1998	2257	2811	3282	4746	5874	6827

PARAMETERS FOR THESE PROCESSES SELECTED SO THAT $\text{VAR}(X(1))=1$. AND $E(X(1))=0.0$. I.E.. $\text{RHO}=1.0$

MEAN WAIT IN THOUSANDS	0.20	0.40	0.60	0.80	1.00	1.50	2.00	3.00	4.00	6.00	8.00	10.00	20.00	40.00	60.00	80.00
EPOCH:	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---
WTENER	491	602	699	784	861	1029	1173	1419	1628	1980	2279	2543	3583	5056	6189	7144
MD1 QUEUE	400	400	521	621	703	857	1010	1253	1461	1814	2113	2377	3416	4890	6022	6977
ME1 K= 4.00	400	400	508	595	670	836	979	1224	1432	1784	2082	2346	3384	4858	5990	6945
ME1 K= 3.00	400	400	504	589	665	830	973	1217	1425	1776	2075	2338	3377	4850	5982	6936
ME1 K= 2.00	400	400	498	580	655	820	962	1205	1413	1764	2062	2326	3364	4836	5968	6923
ME1 K= 1.50	400	400	493	574	648	812	954	1197	1404	1755	2053	2316	3353	4826	5958	6912
ME1 K= 1.00	400	400	486	565	638	801	942	1184	1390	1741	2038	2301	3338	4810	5942	6896
ME1 K= 0.50	400	400	476	552	623	783	923	1163	1369	1718	2015	2277	3313	4784	5916	6870
ME1 K= 0.20	400	400	468	540	610	767	905	1143	1348	1696	1992	2254	3288	4759	5890	6844
GAMMA INPUT	400	400	460	530	597	752	888	1124	1328	1674	1970	2231	3264	4734	5864	6818

PARAMETERS FOR THESE PROCESSES SELECTED SO THAT $\text{VAR}(X(1))=1$. AND $E(X(1))=0.0$. I.E.. $\text{RHO}=1.0$

MEAN WAIT IN THOUSANDS	0.20	0.40	0.60	0.80	1.00	1.50	2.00	3.00	4.00	6.00	8.00	10.00	20.00	40.00	60.00	80.00
EPOCH:	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---
WTENER	392	530	639	731	814	990	1140	1391	1604	1961	2262	2528	3572	5049	6182	7138
MD1 QUEUE	200	348	470	569	651	822	974	1225	1437	1794	2096	2362	3405	4882	6016	6972
ME1 K= 4.00	200	338	443	535	616	790	939	1196	1408	1764	2065	2331	3374	4850	5983	6939
ME1 K= 3.00	200	335	443	535	616	790	939	1189	1401	1757	2058	2323	3366	4842	5975	6931
ME1 K= 2.00	200	330	435	526	606	780	928	1178	1389	1745	2045	2311	3353	4829	5962	6918
ME1 K= 1.50	200	326	430	519	599	772	920	1169	1380	1735	2036	2301	3343	4818	5952	6907
ME1 K= 1.00	200	321	422	510	589	761	907	1156	1366	1721	2021	2286	3327	4803	5936	6891
ME1 K= 0.50	200	312	410	496	574	743	888	1135	1345	1698	1998	2262	3302	4777	5909	6865
ME1 K= 0.20	200	304	399	483	559	726	870	1115	1324	1676	1975	2239	3278	4751	5883	6838
GAMMA INPUT	200	297	389	471	546	711	853	1096	1304	1655	1953	2216	3253	4726	5858	6812

PARAMETERS FOR THESE PROCESSES SELECTED SO THAT $\text{VAR}(X(1))=1$. AND $E(X(1))=0.0$. I.E.. $\text{RHO}=1.0$

MEAN WAIT IN THOUSANDS	0.20	0.40	0.60	0.80	1.00	1.50	2.00	3.00	4.00	6.00	8.00	10.00	20.00	40.00	60.00	80.00
EPOCH:	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---
WTENER	357	505	618	714	798	977	1128	1382	1596	1954	2257	2523	3568	5046	6180	7136
MD1 QUEUE	181	330	451	551	632	810	962	1215	1429	1788	2090	2356	3402	4880	6014	6970
ME1 K= 4.00	177	316	428	522	606	784	934	1187	1400	1758	2060	2326	3370	4848	5981	6937
ME1 K= 3.00	175	312	423	517	600	777	927	1180	1393	1750	2052	2318	3362	4840	5973	6929
ME1 K= 2.00	173	307	415	508	591	767	917	1168	1381	1738	2040	2306	3349	4826	5960	6916
ME1 K= 1.50	171	302	409	502	583	759	908	1159	1372	1729	2030	2296	3339	4816	5949	6905
ME1 K= 1.00	168	296	401	492	573	748	896	1146	1358	1715	2016	2281	3324	4800	5934	6889
ME1 K= 0.50	164	287	389	478	558	733	877	1126	1337	1692	1992	2257	3299	4774	5907	6863
ME1 K= 0.20	159	278	378	465	543	713	859	1106	1316	1670	1969	2234	3274	4749	5881	6837
GAMMA INPUT	155	270	367	453	530	698	842	1087	1296	1648	1947	2211	3250	4723	5856	6811

RHO = 0.90									
	SCALED INITIAL SERVER LOAD = 4.00					EPOCH	SCALED MEAN WAIT IN THOUSANDTHS		
	0.20	0.40	0.60	0.80	1.00		4.0	5.0	6.0
WIENER	3800	3600	3400	3200	3000	1404	1001	774	545
MDI QUEUE	3800	3600	3400	3200	3000	1396	996	772	546
MEI K= 4.00	3800	3600	3400	3200	3000	1394	995	771	546
MEI K= 3.00	3800	3600	3400	3200	3000	1394	994	771	546
MEI K= 2.00	3800	3600	3400	3200	3000	1393	994	771	546
MEI K= 1.50	3800	3600	3400	3200	3000	1393	994	771	546
MEI K= 1.00	3800	3600	3400	3200	3000	1393	994	771	546
MEI K= 0.50	3800	3600	3400	3200	3000	1392	993	771	546
MEI K= 0.20	3800	3600	3400	3200	3000	1389	992	770	546
GAMMA INPUT	3800	3600	3400	3200	3000	1388	990	770	546

RHO = 0.90									
	SCALED INITIAL SERVER LOAD = 3.00					EPOCH	SCALED MEAN WAIT IN THOUSANDTHS		
	0.20	0.40	0.60	0.80	1.00		4.0	5.0	6.0
WIENER	2800	2600	2400	2200	2000	887	694	599	551
MDI QUEUE	2800	2600	2400	2200	2000	882	692	599	552
MEI K= 4.00	2800	2600	2400	2200	2000	880	692	599	552
MEI K= 3.00	2800	2600	2400	2200	2000	880	692	599	552
MEI K= 2.00	2800	2600	2400	2200	2000	880	691	599	552
MEI K= 1.50	2800	2600	2400	2200	2000	879	691	599	552
MEI K= 1.00	2800	2600	2400	2200	2000	879	691	599	552
MEI K= 0.50	2800	2600	2400	2200	2000	877	691	598	552
MEI K= 0.20	2800	2600	2400	2200	2000	876	690	598	552
GAMMA INPUT	2800	2600	2400	2200	2000	876	690	598	552

RHO = 0.90									
	SCALED INITIAL SERVER LOAD = 2.00					EPOCH	SCALED MEAN WAIT IN THOUSANDTHS		
	0.20	0.40	0.60	0.80	1.00		4.0	5.0	6.0
WIENER	1800	1600	1418	1258	1126	609	550	524	512
MDI QUEUE	1800	1600	1413	1250	1116	606	550	524	512
MEI K= 4.00	1800	1600	1412	1249	1114	606	549	524	512
MEI K= 3.00	1800	1600	1412	1248	1114	606	549	524	512
MEI K= 2.00	1800	1600	1411	1248	1113	605	549	524	512
MEI K= 1.50	1800	1600	1411	1247	1112	605	549	524	512
MEI K= 1.00	1800	1600	1411	1246	1111	605	549	524	512
MEI K= 0.50	1800	1600	1410	1245	1110	605	549	524	512
MEI K= 0.20	1800	1600	1410	1244	1109	604	549	524	512
GAMMA INPUT	1800	1600	1409	1243	1107	604	549	524	512

PHC = 0.90																
	SCALED INITIAL SERVER LOAD = 1.00				EPOCH	SCALED MEAN WAIT IN THOUSANDTHS										
	0.05	0.10	0.15	0.20		0.25	0.30	0.35	0.40	0.50	0.6	0.8	1.0	1.5	2.0	3.0
WIENER	950	500	853	812	775	744	717	654	657	629	590	565	532	517	506	501
MD1 QUEUE	950	900	851	806	767	734	707	683	645	618	580	556	526	514	504	501
ME1 K = 4.00	950	900	851	805	766	733	705	681	643	615	578	555	525	513	504	501
ME1 K = 3.00	950	900	851	805	766	732	704	680	643	615	577	554	525	513	504	500
ME1 K = 2.00	950	900	851	805	765	732	703	679	642	613	577	554	524	512	504	500
ME1 K = 1.50	950	900	850	805	765	731	703	679	641	613	576	553	524	512	504	500
ME1 K = 1.00	950	900	850	804	764	730	702	678	640	612	575	552	524	512	503	500
ME1 K = 0.50	950	900	850	804	763	729	700	676	638	610	573	551	523	511	503	500
ME1 K = 0.20	950	900	850	803	762	728	699	674	636	608	572	549	522	511	503	500
GAMMA INPUT	950	900	850	803	761	726	697	672	634	606	570	548	521	510	503	500

RHO = 0.90															
	SCALED INITIAL SERVER LOAD = 0.80				EPOCH	SCALED MEAN WAIT IN THOUSANDTHS									
	0.05	0.10	0.15	0.20		0.25	0.30	0.35	0.40	0.6	0.8	1.0	1.5	2.0	3.0
WIENER	750	703	663	632	608	589	574	561	543	531	516	508	500	498	499
MD1 QUEUE	750	703	657	623	597	577	561	549	532	520	505	500	495	495	499
ME1 K = 4.00	750	700	656	621	595	575	559	547	529	518	505	498	494	496	499
ME1 K = 3.00	750	700	656	621	594	574	559	546	529	517	504	498	494	496	499
ME1 K = 2.00	750	700	655	620	593	573	558	545	528	516	503	497	493	496	499
ME1 K = 1.50	750	700	655	620	593	573	557	545	527	516	503	497	493	496	499
ME1 K = 1.00	750	700	655	619	592	571	556	543	526	515	502	496	492	493	499
ME1 K = 0.50	750	700	654	618	590	570	554	541	524	513	501	495	491	492	499
ME1 K = 0.20	750	700	654	617	589	568	552	540	522	511	499	494	491	495	499
GAMMA INPUT	750	700	653	616	587	566	550	538	520	509	497	492	490	491	498

PHQ = 0.90																
	SCALED INITIAL SERVER LOAD = 0.60				EPOCH	SCALED MEAN WAIT IN THOUSANDTHS										
	0.05	0.10	0.15	0.20		0.25	0.30	0.35	0.40	0.6	0.8	1.0	1.5	2.0	3.0	5.0
WIENER	551	514	491	477	468	463	460	458	457	458	462	467	478	485	493	498
MD1 QUEUE	550	507	480	465	456	450	448	446	446	446	454	460	473	482	491	498
ME1 K = 4.00	550	505	478	462	453	447	445	444	444	446	453	459	472	481	491	498
ME1 K = 3.00	550	505	477	461	452	446	444	443	443	446	452	459	472	481	491	498
ME1 K = 2.00	550	505	477	461	452	446	444	442	442	445	451	458	472	481	491	498
ME1 K = 1.50	550	505	476	460	451	446	443	442	442	445	451	458	471	480	491	498
ME1 K = 1.00	550	504	475	459	450	444	442	441	441	444	450	457	471	480	490	498
ME1 K = 0.50	550	503	474	457	448	442	440	439	439	442	445	456	470	480	490	497
ME1 K = 0.20	550	503	472	455	446	440	438	437	438	440	447	455	469	479	490	497
GAMMA INPUT	550	502	471	454	444	438	436	435	436	439	446	454	466	479	490	497

RHO = 0.90									
	SCALED INITIAL SERVER LOAD = 0.40				EPOCH	SCALED MEAN WAIT IN THOUSANDTHS			
	0.01	0.02	0.03	0.04		0.3	0.4	0.6	1.0
WIENER	390	380	372	365	0.20	374	388	411	442
MD1 QUEUE	390	390	370	360	360	362	377	402	435
ME1 K= 4.00	390	380	370	360	340	359	375	400	434
ME1 K= 3.00	390	380	370	360	344	359	374	400	434
ME1 K= 2.00	390	380	370	360	343	358	373	399	433
ME1 K= 1.50	390	380	370	360	342	357	373	398	433
ME1 K= 1.00	390	380	370	360	341	356	370	397	432
ME1 K= 0.50	390	380	370	360	340	354	370	396	431
ME1 K= 0.20	390	380	370	360	338	352	368	394	430
GAMMA INPUT	390	380	370	360	334	350	366	393	429

RHO = 0.90									
	SCALED INITIAL SERVER LOAD = 0.20				EPOCH	SCALED MEAN WAIT IN THOUSANDTHS			
	0.01	0.02	0.03	0.04		0.3	0.4	0.6	1.0
WIENER	192	193	198	204	0.20	324	349	385	428
MD1 QUEUE	190	180	184	190	290	312	339	377	423
ME1 K= 4.00	190	180	184	190	275	310	337	376	421
ME1 K= 3.00	190	180	181	186	274	310	337	375	421
ME1 K= 2.00	190	180	181	185	273	309	336	374	421
ME1 K= 1.50	190	180	180	185	273	308	335	373	420
ME1 K= 1.00	190	180	179	184	271	307	334	372	420
ME1 K= 0.50	190	180	178	182	269	305	332	370	419
ME1 K= 0.20	190	180	177	180	267	303	331	369	418
GAMMA INPUT	190	180	176	179	266	302	329	367	416

RHO = 0.90									
	SCALED INITIAL SERVER LOAD = 0.0				EPOCH	SCALED MEAN WAIT IN THOUSANDTHS			
	0.01	0.02	0.03	0.04		0.3	0.4	0.6	1.0
WIENER	75	103	124	141	0.20	309	338	378	425
MD1 QUEUE	59	87	108	125	269	298	328	370	419
ME1 K= 4.00	56	84	105	123	257	296	326	368	418
ME1 K= 3.00	55	84	105	122	254	295	325	367	417
ME1 K= 2.00	54	83	104	121	253	294	324	366	416
ME1 K= 1.50	53	82	103	120	252	293	323	365	415
ME1 K= 1.00	52	81	102	119	251	291	322	364	414
ME1 K= 0.50	51	79	100	117	249	289	320	363	413
ME1 K= 0.20	49	77	98	115	247	288	319	362	413
GAMMA INPUT	48	75	96	113	245	288	319	362	413

RHO = 0.80													
	SCALED INITIAL SERVER LOAD =				SCALED				EPOCH	SCALED			
	0.40	0.60	0.80	1.00	1.50	2.00	2.50	3.00		5.0	6.0	7.0	8.0
WIENER	3600	3400	3200	3001	2514	2073	1701	1404	1001	774	649	582	545
MDI QUEUE	3500	3400	3200	3000	2508	2058	1682	1386	989	769	649	583	547
ME1 K= 4.00	3500	3400	3200	3000	2507	2055	1678	1382	987	768	649	583	547
ME1 K= 3.00	3600	3400	3200	3000	2506	2054	1677	1381	986	768	649	583	547
ME1 K= 2.00	3600	3400	3200	3000	2506	2053	1676	1379	985	767	649	583	547
ME1 K= 1.50	3600	3400	3200	3000	2506	2052	1674	1378	985	767	649	583	547
ME1 K= 1.00	3600	3400	3200	3000	2505	2051	1672	1376	983	766	648	584	548
ME1 K= 0.50	3600	3400	3200	3000	2504	2049	1669	1373	981	765	648	584	548
ME1 K= 0.20	3600	3400	3200	3000	2504	2046	1666	1370	979	765	648	584	548
GAMMA INPUT	3600	3400	3200	3000	2503	2044	1663	1367	977	764	648	584	548

RHO = 0.80													
	SCALED INITIAL SERVER LOAD =				SCALED				EPOCH	SCALED			
	0.40	0.60	0.80	1.00	1.50	2.00	2.50	3.00		5.0	6.0	7.0	8.0
WIENER	2600	2400	2203	2013	1596	1278	1045	887	694	599	551	527	514
MDI QUEUE	2600	2400	2201	2006	1578	1258	1032	875	689	598	552	528	515
ME1 K= 4.00	2600	2400	2200	2005	1574	1253	1028	872	688	598	552	528	516
ME1 K= 3.00	2600	2400	2200	2005	1574	1252	1027	871	688	598	552	528	516
ME1 K= 2.00	2600	2400	2200	2004	1572	1251	1026	870	688	598	552	528	516
ME1 K= 1.50	2600	2400	2200	2004	1571	1249	1024	869	687	598	552	528	516
ME1 K= 1.00	2600	2400	2200	2004	1569	1247	1023	868	687	598	552	528	516
ME1 K= 0.50	2600	2400	2200	2003	1567	1241	1020	866	686	597	552	529	516
ME1 K= 0.20	2600	2400	2200	2003	1564	1241	1017	863	685	597	552	529	516
GAMMA INPUT	2600	2400	2200	2002	1561	1237	1014	861	684	597	552	529	516

RHO = 0.80													
	SCALED INITIAL SERVER LOAD =				SCALED				EPOCH	SCALED			
	0.40	0.60	0.80	1.00	1.50	2.00	2.50	3.00		5.0	6.0	7.0	8.0
WIENER	1602	1418	1256	1126	891	750	664	609	550	524	512	506	503
MDI QUEUE	1600	1407	1240	1104	870	735	652	603	549	524	512	506	503
ME1 K= 4.00	1600	1406	1237	1100	866	732	651	602	548	524	512	507	504
ME1 K= 3.00	1600	1406	1236	1098	865	732	651	601	548	524	512	507	504
ME1 K= 2.00	1600	1405	1235	1097	864	730	650	601	548	524	512	507	504
ME1 K= 1.50	1600	1405	1234	1095	862	729	650	600	548	524	512	507	504
ME1 K= 1.00	1600	1404	1232	1093	860	728	649	600	547	524	512	507	504
ME1 K= 0.50	1600	1403	1230	1090	857	725	647	599	547	524	512	507	504
ME1 K= 0.20	1600	1403	1228	1087	857	723	645	598	547	523	512	507	504
GAMMA INPUT	1600	1402	1226	1084	850	720	644	596	546	523	512	507	504

RHO = 0.80									
	0.10	0.15	0.20	0.25	0.30	0.35	0.40	0.50	EPOCH
WIENER	900	853	812	775	744	717	694	657	0.60
MD1 QUEUE	900	850	800	758	723	693	669	631	629
ME1 K= 4.00	900	850	800	756	719	689	663	625	598
ME1 K= 3.00	900	850	800	756	719	688	663	625	597
ME1 K= 2.00	900	850	800	755	717	686	661	622	595
ME1 K= 1.50	900	850	800	754	716	685	660	621	594
ME1 K= 1.00	900	850	800	754	715	683	657	618	591
ME1 K= 0.50	900	850	800	752	713	680	654	615	587
ME1 K= 0.20	900	850	800	752	711	678	651	611	583
GAMMA INPUT	900	850	800	751	709	675	648	607	579
	0.10	0.15	0.20	0.25	0.30	0.35	0.40	0.50	0.60
SCALED INITIAL SERVER LOAD = 1.00									
SCALED	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8
MEAN	506	506	506	506	506	506	506	506	506
WAIT	4.0	4.0	4.0	4.0	4.0	4.0	4.0	4.0	4.0
IN THOUSANDTHS	501	501	501	501	501	501	501	501	501

RHO = 0.80									
	0.10	0.15	0.20	0.25	0.30	0.35	0.40	0.50	EPOCH
WIENER	703	663	632	608	589	574	561	543	531
MD1 QUEUE	700	650	612	583	562	546	534	517	507
ME1 K= 4.00	700	650	609	579	557	541	529	512	502
ME1 K= 3.00	700	650	608	578	556	540	528	511	501
ME1 K= 2.00	700	650	607	577	554	538	525	509	499
ME1 K= 1.50	700	650	606	575	553	536	524	507	497
ME1 K= 1.00	700	650	606	574	551	534	521	505	495
ME1 K= 0.50	700	650	604	571	547	530	517	501	491
ME1 K= 0.20	700	650	603	568	544	526	513	497	487
GAMMA INPUT	700	650	602	566	541	523	510	493	484
	0.10	0.15	0.20	0.25	0.30	0.35	0.40	0.50	0.60
SCALED INITIAL SERVER LOAD = 0.80									
SCALED	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8
MEAN	498	498	498	498	498	498	498	498	498
WAIT	4.0	4.0	4.0	4.0	4.0	4.0	4.0	4.0	4.0
IN THOUSANDTHS	499	499	499	499	499	499	499	499	499

RHO = 0.80									
	0.10	0.15	0.20	0.25	0.30	0.35	0.40	0.50	EPOCH
WIENER	514	491	477	468	462	460	458	457	458
MD1 QUEUE	500	467	449	439	435	432	432	433	436
ME1 K= 4.00	500	463	444	434	429	427	427	428	432
ME1 K= 3.00	500	462	443	433	428	426	425	425	431
ME1 K= 2.00	500	460	441	431	426	424	423	424	429
ME1 K= 1.50	500	460	439	429	424	422	422	424	428
ME1 K= 1.00	500	458	437	427	422	419	419	421	426
ME1 K= 0.50	500	456	434	423	417	415	415	416	422
ME1 K= 0.20	500	455	431	419	413	411	411	414	419
GAMMA INPUT	500	454	428	416	410	407	407	410	415
	0.10	0.15	0.20	0.25	0.30	0.35	0.40	0.50	0.60
SCALED INITIAL SERVER LOAD = 0.60									
SCALED	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8
MEAN	485	485	485	485	485	485	485	485	485
WAIT	3.0	3.0	3.0	3.0	3.0	3.0	3.0	3.0	3.0
IN THOUSANDTHS	498	498	498	498	498	498	498	498	498

RHO = 0.80																
	SCALED INITIAL SERVER LOAD = 0.40				EPOCH	SCALED MEAN WAIT IN THOUSANDTHS										
	0.02	0.04	0.06	0.08		0.10	0.15	0.20	0.30							
WIENER	380	365	357	353	351	354	360	374	388	411	428	442	464	477	490	495
MD1 QUEUE	380	360	340	320	324	323	324	342	363	391	412	428	454	470	486	493
ME1 K = 4.00	380	360	340	320	316	316	320	340	358	386	408	425	452	469	485	493
ME1 K = 3.00	380	360	340	320	315	316	320	340	356	384	406	424	451	468	485	493
ME1 K = 2.00	380	360	340	320	314	312	319	337	354	383	405	422	450	467	485	492
ME1 K = 1.50	380	360	340	320	312	310	316	334	352	381	403	421	449	466	484	492
ME1 K = 1.00	380	360	340	320	310	306	312	330	348	378	401	418	446	465	483	492
ME1 K = 0.50	380	360	340	320	308	302	308	326	344	375	398	416	448	464	483	491
ME1 K = 0.20	380	360	340	320	306	299	304	322	341	372	395	414	444	463	482	491
GAMMA INPUT	380	360	340	320	306	299	304	322	341	372	395	414	444	463	482	491

RHO = 0.80													
	SCALED INITIAL SERVER LOAD = 0.20				EPOCH	SCALED MEAN WAIT IN THOUSANDTHS							
	0.02	0.04	0.06	0.08		0.10	0.15	0.20	0.30				
WIENER	193	204	218	231	349	385	410	385	410				
MD1 QUEUE	180	160	185	197	327	367	395	367	395				
ME1 K = 4.00	180	160	178	192	322	363	392	363	392				
ME1 K = 3.00	180	160	177	190	321	362	391	362	391				
ME1 K = 2.00	180	160	175	188	320	361	390	360	389				
ME1 K = 1.50	180	160	173	186	318	360	389	358	387				
ME1 K = 1.00	180	160	171	184	316	358	387	356	386				
ME1 K = 0.50	180	160	168	180	313	355	384	352	382				
ME1 K = 0.20	180	160	165	177	309	352	382	349	379				
GAMMA INPUT	180	160	163	174	306	349	379	349	379				

RHO = 0.80													
	SCALED INITIAL SERVER LOAD = 0.20				EPOCH	SCALED MEAN WAIT IN THOUSANDTHS							
	0.02	0.04	0.06	0.08		0.10	0.15	0.20	0.30				
WIENER	103	141	167	189	338	378	405	378	405				
MD1 QUEUE	65	108	134	157	316	360	390	360	390				
ME1 K = 4.00	53	102	130	152	312	357	387	357	387				
ME1 K = 3.00	62	100	128	151	311	356	386	356	386				
ME1 K = 2.00	61	99	126	149	309	354	385	354	385				
ME1 K = 1.50	60	97	125	147	308	353	384	353	384				
ME1 K = 1.00	58	95	123	145	306	351	382	351	382				
ME1 K = 0.50	57	92	119	141	303	348	380	348	380				
ME1 K = 0.20	55	89	116	138	299	345	377	345	377				
GAMMA INPUT	53	87	113	135	296	343	375	343	375				

RHO = 0.70																
WIENER																
MD1 QUEUE																
ME1 K = 4.00																
ME1 K = 3.00																
ME1 K = 2.00																
ME1 K = 1.50																
ME1 K = 1.00																
ME1 K = 0.50																
ME1 K = 0.20																
GAMMA INPUT																
0.50	1.00	1.50	2.00	2.50	3.00	4.00	5.00	SCALED	EPOCH	7.0	8.0	9.0	10.0	12.0	14.0	16.0
3500	3001	2514	2073	1701	1404	1001	774	582	649	582	545	525	514	504	501	500
3500	3000	2504	2047	1668	1372	981	766	584	649	584	548	527	516	505	502	501
3500	3000	2503	2042	1661	1365	977	765	584	648	584	548	528	516	506	502	501
3500	3000	2502	2041	1660	1364	976	764	584	648	584	548	528	516	506	502	501
3500	3000	2502	2039	1657	1361	974	763	584	648	584	549	528	517	506	502	501
3500	3000	2502	2038	1655	1359	973	763	584	648	584	549	529	517	506	502	501
3500	3000	2501	2036	1652	1356	971	762	584	648	584	545	529	517	506	502	501
3500	3000	2501	2033	1646	1350	967	760	585	647	585	549	529	517	506	502	501
3500	3000	2501	2030	1641	1345	964	759	585	647	585	549	529	517	506	502	501
3500	3000	2500	2027	1636	1339	960	757	585	646	585	550	530	518	507	502	501

RHO = 0.70																
WIENER																
MD1 QUEUE																
ME1 K = 4.00																
ME1 K = 3.00																
ME1 K = 2.00																
ME1 K = 1.50																
ME1 K = 1.00																
ME1 K = 0.50																
ME1 K = 0.20																
GAMMA INPUT																
0.50	1.00	1.50	2.00	2.50	3.00	4.00	5.00	SCALED	EPOCH	7.0	8.0	9.0	10.0	12.0	14.0	16.0
2500	2013	1596	1278	1049	887	694	599	527	551	527	514	508	504	501	500	500
2500	2002	1565	1243	1015	866	686	597	529	552	529	516	509	505	502	501	500
2500	2001	1559	1235	1013	861	684	597	529	552	529	516	509	505	502	501	500
2500	2001	1558	1234	1011	860	683	597	529	552	529	516	509	505	502	501	500
2500	2001	1555	1231	1009	858	682	597	529	552	529	516	509	505	502	501	500
2500	2001	1554	1228	1007	856	682	596	529	552	529	516	509	505	502	501	500
2500	2000	1551	1225	1004	854	681	596	529	552	529	517	510	506	502	501	500
2500	2000	1547	1219	999	850	679	595	529	552	529	517	510	506	502	501	500
2500	2000	1543	1214	993	846	677	595	529	552	529	517	510	506	502	501	500
2500	2000	1539	1208	988	842	675	594	530	552	530	517	510	506	502	501	500

RHO = 0.70																
WIENER																
MD1 QUEUE																
ME1 K = 4.00																
ME1 K = 3.00																
ME1 K = 2.00																
ME1 K = 1.50																
ME1 K = 1.00																
ME1 K = 0.50																
ME1 K = 0.20																
GAMMA INPUT																
0.50	1.00	1.50	2.00	2.50	3.00	4.00	5.00	SCALED	EPOCH	7.0	8.0	9.0	10.0	12.0	14.0	16.0
1507	1126	891	750	664	609	550	524	506	512	506	503	502	501	500	500	500
1500	1088	856	725	647	599	547	523	507	512	507	504	502	501	500	500	500
1500	1081	849	720	644	596	546	523	507	512	507	504	502	501	500	500	500
1500	1079	847	718	643	596	546	523	507	512	507	504	502	501	500	500	500
1500	1076	844	716	641	595	546	523	507	512	507	504	502	501	500	500	500
1500	1074	841	714	640	594	545	523	507	512	507	504	502	501	500	500	500
1500	1071	838	712	638	593	545	523	507	512	507	504	502	501	500	500	500
1500	1066	832	707	635	591	544	522	507	512	507	504	502	501	500	500	500
1500	1061	826	703	632	589	543	522	507	512	507	504	502	501	500	500	500
1500	1056	821	698	629	587	542	522	506	512	506	504	502	501	500	500	500

RHO = 0.70									
SCALED INITIAL SERVER LOAD = 0.40									
	0.02	0.04	0.06	0.08	0.10	0.15	0.20	SCALED	EPOCH
	---	---	---	---	---	---	---	0.30	0.40
WIENER	380	365	357	353	351	354	360	374	388
MD1 QUEUE	380	360	340	320	300	298	313	329	346
ME1 K = 4.00	380	360	340	320	300	290	299	319	339
ME1 K = 3.00	380	360	340	320	300	288	297	317	337
ME1 K = 2.00	380	360	340	320	300	285	294	314	334
ME1 K = 1.50	380	360	340	320	300	283	291	311	331
ME1 K = 1.00	380	360	340	320	300	280	287	308	328
ME1 K = 0.50	380	360	340	320	300	276	281	301	322
ME1 K = 0.20	380	360	340	320	300	272	276	296	316
GAMMA INPUT	380	360	340	320	300	269	271	290	311
SCALED MEAN WAIT IN THOUSANDTHS									
	0.02	0.04	0.06	0.08	0.10	0.15	0.20	0.8	1.0
	---	---	---	---	---	---	---	---	---
WIENER	495	495	495	495	495	495	495	495	495
MD1 QUEUE	495	495	495	495	495	495	495	495	495
ME1 K = 4.00	495	495	495	495	495	495	495	495	495
ME1 K = 3.00	495	495	495	495	495	495	495	495	495
ME1 K = 2.00	495	495	495	495	495	495	495	495	495
ME1 K = 1.50	495	495	495	495	495	495	495	495	495
ME1 K = 1.00	495	495	495	495	495	495	495	495	495
ME1 K = 0.50	495	495	495	495	495	495	495	495	495
ME1 K = 0.20	495	495	495	495	495	495	495	495	495
GAMMA INPUT	495	495	495	495	495	495	495	495	495

RHO = 0.70									
SCALED INITIAL SERVER LOAD = 0.20									
	0.02	0.04	0.06	0.08	0.10	0.15	0.20	SCALED	EPOCH
	---	---	---	---	---	---	---	0.30	0.40
WIENER	193	204	218	231	243	269	290	324	349
MD1 QUEUE	180	160	140	166	186	221	240	282	312
ME1 K = 4.00	180	160	140	161	178	209	234	274	304
ME1 K = 3.00	180	160	140	160	176	207	233	272	303
ME1 K = 2.00	180	160	140	158	173	204	229	269	300
ME1 K = 1.50	180	160	140	157	171	201	227	267	297
ME1 K = 1.00	180	160	140	154	168	198	223	263	294
ME1 K = 0.50	180	160	140	151	163	192	217	258	289
ME1 K = 0.20	180	160	140	148	159	188	212	252	283
GAMMA INPUT	180	160	140	145	156	183	207	247	278
SCALED MEAN WAIT IN THOUSANDTHS									
	0.02	0.04	0.06	0.08	0.10	0.15	0.20	0.8	1.0
	---	---	---	---	---	---	---	---	---
WIENER	486	486	486	486	486	486	486	486	486
MD1 QUEUE	486	486	486	486	486	486	486	486	486
ME1 K = 4.00	486	486	486	486	486	486	486	486	486
ME1 K = 3.00	486	486	486	486	486	486	486	486	486
ME1 K = 2.00	486	486	486	486	486	486	486	486	486
ME1 K = 1.50	486	486	486	486	486	486	486	486	486
ME1 K = 1.00	486	486	486	486	486	486	486	486	486
ME1 K = 0.50	486	486	486	486	486	486	486	486	486
ME1 K = 0.20	486	486	486	486	486	486	486	486	486
GAMMA INPUT	486	486	486	486	486	486	486	486	486

RHO = 0.70									
SCALED INITIAL SERVER LOAD = 0.0									
	0.02	0.04	0.06	0.08	0.10	0.15	0.20	SCALED	EPOCH
	---	---	---	---	---	---	---	0.30	0.40
WIENER	103	141	167	189	206	242	269	309	338
MD1 QUEUE	43	80	111	136	157	194	224	268	301
ME1 K = 4.00	42	77	105	128	148	186	216	261	294
ME1 K = 3.00	42	76	104	126	146	184	214	259	293
ME1 K = 2.00	42	75	101	124	143	181	211	256	290
ME1 K = 1.50	41	74	100	122	141	179	209	254	288
ME1 K = 1.00	41	72	98	119	138	176	205	251	285
ME1 K = 0.50	39	70	94	115	133	171	200	245	279
ME1 K = 0.20	38	67	91	111	129	166	195	240	274
GAMMA INPUT	37	65	88	108	125	161	190	235	269
SCALED MEAN WAIT IN THOUSANDTHS									
	0.02	0.04	0.06	0.08	0.10	0.15	0.20	0.8	1.0
	---	---	---	---	---	---	---	---	---
WIENER	488	488	488	488	488	488	488	488	488
MD1 QUEUE	488	488	488	488	488	488	488	488	488
ME1 K = 4.00	488	488	488	488	488	488	488	488	488
ME1 K = 3.00	488	488	488	488	488	488	488	488	488
ME1 K = 2.00	488	488	488	488	488	488	488	488	488
ME1 K = 1.50	488	488	488	488	488	488	488	488	488
ME1 K = 1.00	488	488	488	488	488	488	488	488	488
ME1 K = 0.50	488	488	488	488	488	488	488	488	488
ME1 K = 0.20	488	488	488	488	488	488	488	488	488
GAMMA INPUT	488	488	488	488	488	488	488	488	488

RHO = 0.60													
	SCALED INITIAL SERVER LOAD = 4.00					EPOCH	SCALED MEAN WAIT IN THOUSANDTHS			SCALED MEAN WAIT IN THOUSANDTHS			
	C.50	1.00	1.50	2.00	2.50		6.0	7.0	8.0	10.0	12.0	15.0	20.0
WIENER	3500	3001	2514	2073	1701	5.00	649	582	545	514	504	501	500
MD1 QUEUE	3500	3000	2500	2032	1648	5.00	774	585	549	517	506	501	500
ME1 K= 4.00	3500	3000	2500	2026	1638	5.00	762	585	550	517	506	501	500
ME1 K= 3.00	3500	3000	2500	2025	1636	5.00	759	585	550	518	506	502	500
ME1 K= 2.00	3500	3000	2500	2022	1632	5.00	757	585	550	518	507	502	500
ME1 K= 1.50	3500	3000	2500	2021	1629	5.00	756	585	550	518	507	502	500
ME1 K= 1.00	3500	3000	2500	2018	1624	5.00	755	585	550	518	507	502	500
ME1 K= 0.50	3500	3000	2500	2015	1615	5.00	752	585	551	519	507	502	500
ME1 K= 0.20	3500	3000	2500	2012	1609	5.00	749	585	551	519	507	502	500
GAMMA INPUT	3500	3000	2500	2009	1602	5.00	746	585	551	519	508	502	500

RHO = 0.60													
	SCALED INITIAL SERVER LOAD = 3.00					EPOCH	SCALED MEAN WAIT IN THOUSANDTHS			SCALED MEAN WAIT IN THOUSANDTHS			
	0.50	1.00	1.50	2.00	2.50		6.0	7.0	8.0	10.0	12.0	15.0	20.0
WIENER	2500	2013	1596	1278	1049	5.00	551	527	514	504	501	500	500
MD1 QUEUE	2500	2000	1547	1222	1002	5.00	553	529	517	506	502	500	500
ME1 K= 4.00	2500	2000	1538	1211	993	5.00	552	530	517	506	502	500	500
ME1 K= 3.00	2500	2000	1536	1208	990	5.00	552	530	517	506	502	500	500
ME1 K= 2.00	2500	2000	1533	1204	986	5.00	552	530	517	506	502	501	500
ME1 K= 1.50	2500	2000	1531	1200	983	5.00	552	530	517	506	502	501	500
ME1 K= 1.00	2500	2000	1528	1195	978	5.00	552	530	517	506	502	501	500
ME1 K= 0.50	2500	2000	1523	1186	970	5.00	552	530	518	506	502	501	500
ME1 K= 0.20	2500	2000	1519	1178	962	5.00	551	530	518	507	503	501	500
GAMMA INPUT	2500	2000	1515	1170	954	5.00	551	530	518	507	503	501	500

RHO = 0.60													
	SCALED INITIAL SERVER LOAD = 2.00					EPOCH	SCALED MEAN WAIT IN THOUSANDTHS			SCALED MEAN WAIT IN THOUSANDTHS			
	0.50	1.00	1.50	2.00	2.50		6.0	7.0	8.0	10.0	12.0	15.0	20.0
WIENER	1507	1126	891	750	664	5.00	512	506	503	501	500	500	500
MD1 QUEUE	1500	1066	836	711	638	5.00	512	507	504	501	500	500	500
ME1 K= 4.00	1500	1055	825	703	632	5.00	512	507	504	501	500	500	500
ME1 K= 3.00	1500	1053	822	701	631	5.00	512	507	504	501	500	500	500
ME1 K= 2.00	1500	1049	817	697	629	5.00	512	507	504	501	500	500	500
ME1 K= 1.50	1500	1046	814	694	627	5.00	512	507	504	501	500	500	500
ME1 K= 1.00	1500	1041	808	690	624	5.00	511	506	504	501	500	500	500
ME1 K= 0.50	1500	1035	799	683	619	5.00	511	506	504	501	500	500	500
ME1 K= 0.20	1500	1029	791	676	614	5.00	511	506	504	501	500	500	500
GAMMA INPUT	1500	1025	782	670	609	5.00	510	506	503	501	500	500	500

RHO = 0.60									
SCALED INITIAL SERVER LOAD = 1.00									
	0.20	0.30	0.40	0.50	0.60	0.80	1.00	1.50	EPOCH
WIENER	812	744	694	657	629	590	565	532	2.00
MD1 QUEUE	800	700	600	592	561	533	515	500	517
ME1 K= 4.00	800	700	600	573	548	519	505	493	496
ME1 K= 3.00	800	700	600	570	544	517	503	491	491
ME1 K= 2.00	800	700	600	565	539	512	499	489	488
ME1 K= 1.50	800	700	600	561	535	508	496	486	486
ME1 K= 1.00	800	700	600	556	530	503	491	483	483
ME1 K= 0.50	800	700	600	548	521	494	483	477	480
ME1 K= 0.20	800	700	600	542	513	486	475	471	476
GAMMA INPUT	800	700	600	536	506	478	468	466	471
SCALED MEAN WAIT IN THOUSANDTHS									
	0.20	0.30	0.40	0.50	0.60	0.80	1.00	1.50	EPOCH
WIENER	501	501	501	506	502	502	506	506	510
MD1 QUEUE	500	500	500	496	497	497	496	496	495
ME1 K= 4.00	500	500	500	493	496	496	493	493	492
ME1 K= 3.00	500	500	500	491	495	495	491	491	491
ME1 K= 2.00	500	500	500	492	495	495	492	492	490
ME1 K= 1.50	500	500	500	491	494	494	491	491	489
ME1 K= 1.00	500	500	500	489	494	494	489	489	488
ME1 K= 0.50	500	500	500	487	496	496	487	487	484
ME1 K= 0.20	500	500	500	480	495	495	485	485	480
GAMMA INPUT	498	493	493	477	482	482	477	477	471

RHO = 0.60									
SCALED INITIAL SERVER LOAD = 0.80									
	0.20	0.30	0.40	0.50	0.60	0.80	1.00	1.50	EPOCH
WIENER	632	589	561	543	531	516	508	500	498
MD1 QUEUE	600	500	489	479	462	464	463	471	478
ME1 K= 4.00	600	500	474	461	455	452	455	465	474
ME1 K= 3.00	600	500	471	458	452	450	452	463	473
ME1 K= 2.00	600	500	465	453	447	445	449	461	471
ME1 K= 1.50	600	500	462	449	443	442	446	458	469
ME1 K= 1.00	600	500	457	443	438	437	441	455	467
ME1 K= 0.50	600	500	449	435	429	429	434	450	463
ME1 K= 0.20	600	500	443	427	421	421	427	445	459
GAMMA INPUT	600	500	437	419	413	414	421	440	455
SCALED MEAN WAIT IN THOUSANDTHS									
	0.20	0.30	0.40	0.50	0.60	0.80	1.00	1.50	EPOCH
WIENER	499	499	499	498	498	499	498	498	498
MD1 QUEUE	499	499	499	496	496	496	496	496	496
ME1 K= 4.00	499	499	499	496	496	496	496	496	496
ME1 K= 3.00	499	499	499	495	495	495	495	495	495
ME1 K= 2.00	499	499	499	495	495	495	495	495	495
ME1 K= 1.50	499	499	499	495	495	495	495	495	495
ME1 K= 1.00	499	499	499	494	494	494	494	494	494
ME1 K= 0.50	499	499	499	493	493	493	493	493	493
ME1 K= 0.20	499	499	499	492	492	492	492	492	492
GAMMA INPUT	498	491	491	485	485	485	485	485	485

RHO = 0.60									
SCALED INITIAL SERVER LOAD = 0.60									
	0.20	0.30	0.40	0.50	0.60	0.80	1.00	1.50	EPOCH
WIENER	477	463	458	457	458	462	467	478	485
MD1 QUEUE	400	382	396	389	402	414	428	451	466
ME1 K= 4.00	400	371	376	382	390	406	420	445	462
ME1 K= 3.00	400	368	373	379	387	403	418	444	461
ME1 K= 2.00	400	363	367	375	383	400	414	441	459
ME1 K= 1.50	400	360	363	371	380	397	411	439	458
ME1 K= 1.00	400	355	358	366	374	392	408	436	455
ME1 K= 0.50	400	348	350	357	366	385	401	432	452
ME1 K= 0.20	400	343	342	349	359	378	395	427	448
GAMMA INPUT	400	337	335	342	351	371	388	422	444
SCALED MEAN WAIT IN THOUSANDTHS									
	0.20	0.30	0.40	0.50	0.60	0.80	1.00	1.50	EPOCH
WIENER	498	498	498	498	498	498	498	498	498
MD1 QUEUE	498	498	498	498	498	498	498	498	498
ME1 K= 4.00	498	498	498	498	498	498	498	498	498
ME1 K= 3.00	498	498	498	498	498	498	498	498	498
ME1 K= 2.00	498	498	498	498	498	498	498	498	498
ME1 K= 1.50	498	498	498	498	498	498	498	498	498
ME1 K= 1.00	498	498	498	498	498	498	498	498	498
ME1 K= 0.50	498	498	498	498	498	498	498	498	498
ME1 K= 0.20	498	498	498	498	498	498	498	498	498
GAMMA INPUT	498	490	490	483	483	483	483	483	483

RHO = 0.60													
SCALED INITIAL SERVER LOAD = 0.40													
	0.05	0.10	0.15	0.20	0.30	0.40	0.50	0.60	EPOCH	1.0	1.2	1.5	2.0
WIENER	360	351	354	360	374	388	400	411	428	442	452	464	477
MD1 QUEUE	350	300	250	267	310	323	342	361	385	405	421	439	459
ME1 K= 4.00	350	300	250	260	291	313	332	349	377	398	414	433	454
ME1 K= 3.00	350	300	250	259	288	310	330	347	375	396	412	432	453
ME1 K= 2.00	350	300	250	255	283	306	326	343	371	393	410	429	452
ME1 K= 1.50	350	300	250	253	275	302	322	340	368	386	404	425	448
ME1 K= 1.00	350	300	250	249	274	297	318	335	364	386	404	425	448
ME1 K= 0.50	350	300	250	244	266	289	310	328	357	380	404	427	448
ME1 K= 0.20	350	300	250	239	260	282	303	321	351	374	393	415	441
GAMMA INPUT	350	300	250	235	253	275	296	314	345	369	386	411	437
SCALED MEAN WAIT IN THOUSANDTHS													
	3.0	4.0	5.0										
	450	495	498										
	480	489	494										
	477	486	493										
	477	488	493										
	476	487	493										
	475	486	492										
	474	486	492										
	471	484	491										
	469	483	490										
	467	481	485										

RHO = 0.60													
SCALED INITIAL SERVER LOAD = 0.20													
	0.05	0.10	0.15	0.20	0.30	0.40	0.50	0.60	EPOCH	1.0	1.2	1.5	2.0
WIENER	211	243	265	290	324	349	365	385	410	428	442	457	473
MD1 QUEUE	150	141	185	220	262	290	318	337	370	393	412	432	455
ME1 K= 4.00	150	139	177	206	249	281	307	328	362	386	405	427	451
ME1 K= 3.00	150	138	175	203	246	279	305	326	360	384	404	425	450
ME1 K= 2.00	150	137	172	199	242	275	301	322	356	381	401	423	448
ME1 K= 1.50	150	136	169	196	239	271	298	319	354	379	399	421	446
ME1 K= 1.00	150	134	166	192	234	267	293	315	350	376	396	419	444
ME1 K= 0.50	150	131	160	185	227	259	286	308	343	370	390	414	441
ME1 K= 0.20	150	129	155	180	220	253	275	302	337	364	385	409	437
GAMMA INPUT	150	126	151	174	214	246	273	295	331	358	380	405	433
SCALED MEAN WAIT IN THOUSANDTHS													
	3.0	4.0	5.0										
	488	494	497										
	476	489	494										
	475	487	493										
	474	486	492										
	473	486	492										
	472	485	492										
	470	483	491										
	467	482	490										
	465	480	489										

RHO = 0.60													
SCALED INITIAL SERVER LOAD = 0.0													
	0.05	0.10	0.15	0.20	0.30	0.40	0.50	0.60	EPOCH	1.0	1.2	1.5	2.0
WIENER	155	206	242	269	309	338	360	378	405	425	435	455	472
MD1 QUEUE	68	124	168	203	247	282	310	331	365	390	409	430	454
ME1 K= 4.00	57	118	158	190	237	272	300	322	357	383	403	425	449
ME1 K= 3.00	56	117	156	187	234	270	297	320	355	381	401	424	448
ME1 K= 2.00	55	114	153	183	230	266	294	316	352	378	395	421	447
ME1 K= 1.50	54	113	150	180	227	263	291	314	345	376	396	419	445
ME1 K= 1.00	53	110	147	176	223	258	286	309	345	372	393	417	443
ME1 K= 0.50	61	106	141	170	216	251	279	303	339	367	388	412	440
ME1 K= 0.20	60	102	137	165	210	245	273	296	333	361	383	408	436
GAMMA INPUT	58	99	132	160	204	239	267	290	327	356	378	403	432
SCALED MEAN WAIT IN THOUSANDTHS													
	3.0	4.0	5.0										
	488	494	497										
	478	488	494										
	475	487	493										
	474	486	492										
	473	485	492										
	471	483	491										
	469	482	490										
	467	482	490										
	464	480	486										

RHO = 0.50									
SCALED INITIAL SERVER LOAD = 4.00									
	0.50	1.00	1.50	2.00	2.50	3.00	3.50	4.00	EPOCH
WIENER	3500	3001	2514	2073	1701	1404	1175	1001	4.50
MD1 QUEUE	3500	3000	2500	2000	1621	1327	1113	956	870
ME1 K= 4.00	3500	3000	2500	2000	1621	1312	1100	946	834
ME1 K= 3.00	3500	3000	2500	2000	1602	1308	1090	943	832
ME1 K= 2.00	3500	3000	2500	2000	1596	1302	1091	939	829
ME1 K= 1.50	3500	3000	2500	2000	1591	1296	1086	936	826
ME1 K= 1.00	3500	3000	2500	2000	1584	1289	1080	930	823
ME1 K= 0.50	3500	3000	2500	2000	1574	1276	1068	922	816
ME1 K= 0.20	3500	3000	2500	2000	1565	1264	1058	913	810
GAMMA INPUT	3500	3000	2500	2000	1557	1253	1047	904	803
SCALED MEAN WAIT IN THOUSANDTHS									
	5.0	6.0	7.0	8.0	10.0	15.0	20.0		
	774	649	582	545	514	501	500		
	756	647	586	551	518	502	500		
	752	645	586	551	519	502	500		
	750	645	586	551	519	502	500		
	748	644	586	552	520	502	500		
	747	644	585	552	520	502	500		
	744	643	585	552	520	502	500		
	739	641	585	552	520	502	500		
	735	639	584	552	521	503	500		
	730	636	583	552	521	503	500		

RHO = 0.50									
SCALED INITIAL SERVER LOAD = 3.00									
	0.50	1.00	1.50	2.00	2.50	3.00	3.50	4.00	EPOCH
WIENER	2500	2013	1596	1278	1049	867	774	654	639
MD1 QUEUE	2500	2000	1500	1192	980	837	741	675	628
ME1 K= 4.00	2500	2000	1500	1176	961	826	732	669	624
ME1 K= 3.00	2500	2000	1500	1171	955	819	728	666	622
ME1 K= 2.00	2500	2000	1500	1165	955	819	728	666	622
ME1 K= 1.50	2500	2000	1500	1159	950	815	725	664	621
ME1 K= 1.00	2500	2000	1500	1152	942	809	721	661	619
ME1 K= 0.50	2500	2000	1500	1140	930	799	714	656	615
ME1 K= 0.20	2500	2000	1500	1128	915	790	707	651	612
GAMMA INPUT	2500	2000	1500	1118	908	781	700	645	608
SCALED MEAN WAIT IN THOUSANDTHS									
	5.0	6.0	7.0	8.0	10.0	15.0	20.0		
	599	551	527	514	504	500	500		
	594	553	530	518	506	501	500		
	592	552	530	518	507	501	500		
	591	552	530	518	507	501	500		
	590	551	530	518	507	501	500		
	589	551	530	518	507	501	500		
	586	550	529	518	507	501	500		
	584	549	529	518	507	501	500		
	581	547	529	518	507	501	500		

RHO = 0.50									
SCALED INITIAL SERVER LOAD = 2.00									
	0.50	1.00	1.50	2.00	2.50	3.00	3.50	4.00	EPOCH
WIENER	1500	1126	891	750	664	609	574	550	535
MD1 QUEUE	1500	1000	804	693	625	585	559	542	530
ME1 K= 4.00	1500	1000	791	680	617	579	555	539	528
ME1 K= 3.00	1500	1000	787	676	615	577	554	538	527
ME1 K= 2.00	1500	1000	780	671	611	575	552	537	526
ME1 K= 1.50	1500	1000	775	667	608	573	550	536	526
ME1 K= 1.00	1500	1000	767	660	603	569	548	534	524
ME1 K= 0.50	1500	1000	754	650	596	564	544	531	522
ME1 K= 0.20	1500	1000	743	640	588	558	540	528	520
GAMMA INPUT	1500	1000	731	630	580	552	535	525	517
SCALED MEAN WAIT IN THOUSANDTHS									
	5.0	6.0	7.0	8.0	10.0	15.0	20.0		
	524	512	506	503	501	500	500		
	522	512	507	504	501	500	500		
	520	511	506	504	501	500	500		
	519	511	506	504	501	500	500		
	519	510	505	503	501	500	500		
	518	510	505	503	501	500	500		
	516	509	505	503	501	500	500		
	514	508	505	503	501	500	500		
	513	507	504	502	501	500	500		

RHO = 0.50													
	SCALED INITIAL SERVER LOAD = 1.00				EPOCH	SCALED MEAN WAIT IN THOUSANDTHS							
	0.30	0.40	0.50	0.60		0.80	1.00	1.50	2.00				
WTENER	744	694	557	529	517	506	502	501	500				
MD1 QUEUE	700	600	500	515	486	490	494	496	500				
ME1 K = 4.00	700	600	500	501	478	486	492	495	500				
ME1 K = 3.00	700	600	500	497	476	485	491	495	500				
ME1 K = 2.00	700	600	500	491	473	479	490	494	500				
ME1 K = 1.50	700	600	500	486	470	477	489	493	499				
ME1 K = 1.00	700	600	500	480	466	474	482	492	499				
ME1 K = 0.50	700	600	500	470	460	469	476	482	495				
ME1 K = 0.20	700	600	500	461	454	464	472	483	493				
GAMMA INPUT	700	600	500	454	447	459	468	487	492				

RHO = 0.50													
	SCALED INITIAL SERVER LOAD = 0.80				EPOCH	SCALED MEAN WAIT IN THOUSANDTHS							
	0.30	0.40	0.50	0.60		0.80	1.00	1.50	2.00				
WIENER	589	561	543	531	498	498	498	499	500	500	500	500	500
MD1 QUEUE	500	400	422	443	477	477	483	490	494	497	498	499	500
ME1 K = 4.00	500	400	414	423	468	471	479	488	493	496	497	499	500
ME1 K = 3.00	500	400	411	419	460	470	478	487	493	496	497	498	499
ME1 K = 2.00	500	400	405	407	457	468	476	486	492	495	497	498	499
ME1 K = 1.50	500	400	400	401	441	466	474	485	491	495	497	498	499
ME1 K = 1.00	500	400	394	401	438	454	472	484	490	494	496	498	499
ME1 K = 0.50	500	400	385	383	425	444	468	481	489	493	496	497	499
ME1 K = 0.20	500	400	377	380	412	438	464	479	487	492	495	497	499
GAMMA INPUT	500	400	369	371	411	432	460	476	485	491	494	496	498

RHO = 0.50																
	SCALED INITIAL SERVER LOAD = 0.60				EPOCH	SCALED MEAN WAIT IN THOUSANDTHS										
	0.30	0.40	0.50	0.60		0.80	1.00	1.50	2.00	3.0	4.0	5.0	6.0	7.0	8.0	10.0
WTENER	453	458	457	458	462	467	478	485	490	493	496	498	499	500	500	500
MD1 QUEUE	300	341	368	385	387	411	438	456	469	478	488	493	496	498	499	499
ME1 K = 4.00	300	329	347	359	379	397	429	450	464	474	485	492	495	497	498	499
ME1 K = 3.00	300	326	342	355	376	394	427	448	463	473	485	491	495	497	498	499
ME1 K = 2.00	300	321	336	348	370	389	420	445	460	471	484	490	494	497	498	499
ME1 K = 1.50	300	317	331	343	366	385	420	443	458	469	483	490	494	496	498	499
ME1 K = 1.00	300	311	324	337	360	379	416	439	456	467	481	489	493	496	497	499
ME1 K = 0.50	300	303	314	326	350	370	408	434	451	463	479	489	492	495	497	499
ME1 K = 0.20	300	296	305	317	341	352	401	428	446	460	476	486	491	494	496	498
GAMMA INPUT	300	289	297	309	312	354	394	422	442	456	474	484	490	493	496	498

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20. ABSTRACT (Continue on reverse side if necessary and identify by block number) This research provides numerical results for time-dependent expected server load (mean virtual waiting time) in single-server queues with Poisson arrivals and gamma distributed service times. The results are presented in tabular form to facilitate their use by practitioners involved in the study of operating systems. <i>next page</i> continued		

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ABSTRACT (continued)

The server load process is expressed as a net input process (having stationary, independent increments) modified by a reflecting barrier at the origin. In queueing systems the net input process is compound Poisson. Two other choices of net input processes, the Wiener process (or Brownian motion) and the gamma process, provide approximations for the queueing process. This research considers groups of server load processes whose parameters are selected so that the first and second moments of their net input processes are matched. An existing Laplace transform expression is employed to obtain transient expected server load at specified epochs.

The Laplace transform is inverted numerically using an existing technique. The inversion technique is first applied to test functions whose exact inverses are known, and the results are also compared with queueing results obtained by other methods. These investigations indicate that the numerical results for expected server load have four to six significant digits when the approximation is based upon thirty four values of the Laplace transform function. The computer programs are coded in FORTRAN IV using extended double precision, and complete documentation is included.

The numerical results for expected server load are tabulated in scaled form. Each of the 93 tables includes results for a specific combination of traffic intensity parameter (twelve values, ranging from .5 through 1.0 up to 2.0) and initial server load (either six or nine values, ranging from 0 to 4 in scaled form). An individual tables has results for the Wiener process, the gamma input process, and eight queues with different service time distributions; each of the ten processes is evaluated at sixteen epochs. Step-by-step procedures for using the tables are included, and several sample problems are presented.

The tabulated results allow a comprehensive study of the error associated with using the Wiener process as an approximation of server load in queues. This study confirms that the Wiener process is always an upper bound and that the approximation is best for queues with a traffic intensity parameter near unity. The scaled results also indicate that the gamma input process and queueing process with deterministic service times provide tight lower and upper bounds, respectively, for expected server load in all queues with Poisson arrivals and gamma distributed service times.

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